

AI-BASED IDENTIFICATION OF BINDER AND FIBERS IN 3D IMAGES OF NONWOVEN

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Math2Market

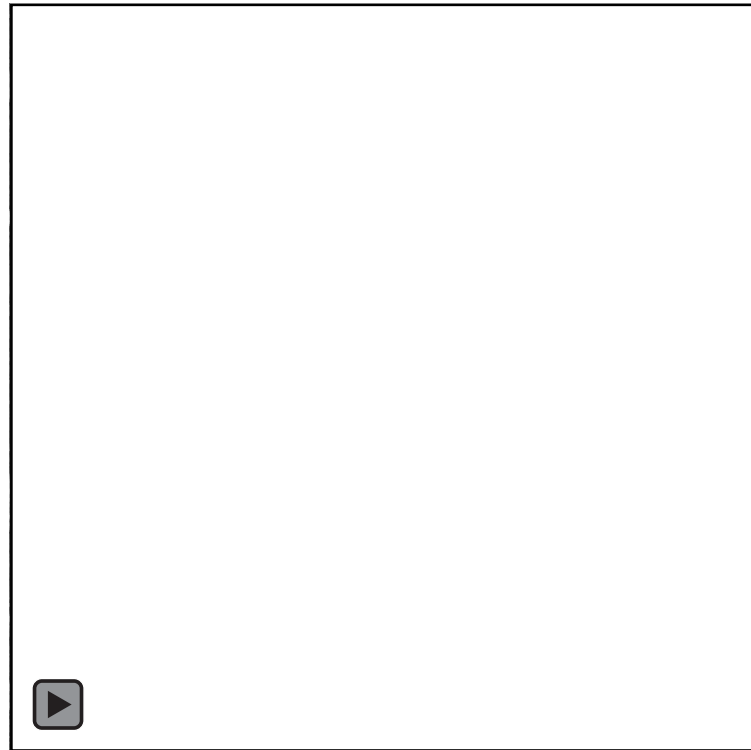
Hans Hagen, TU Kaiserslautern

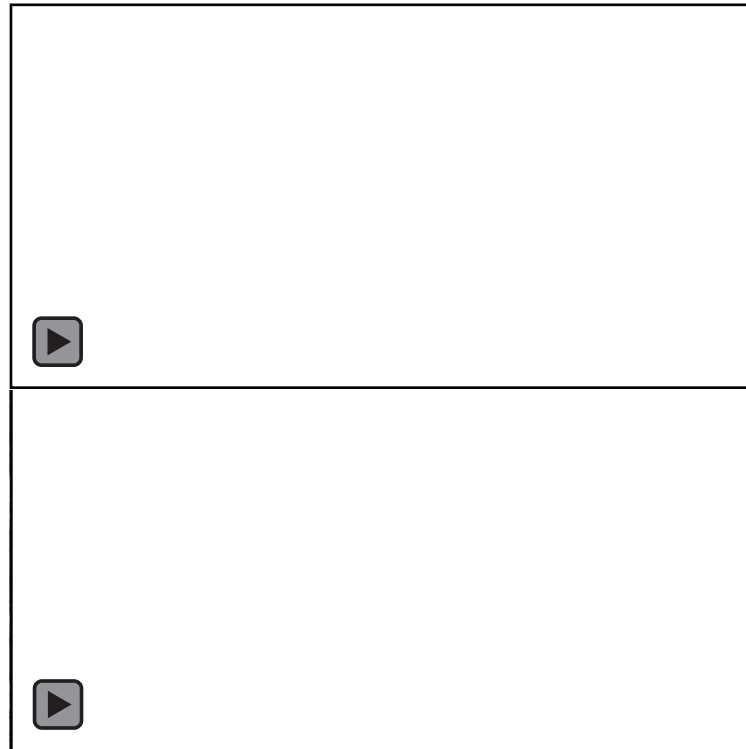
OUR PURPOSE AS A COMPANY RATHER THAN AN ACADEMIC INSTITUTION

- In everything that follows,
- Our intention is to make the technologies available to you, the audience
 - to speed up your modelling of processes and materials
 - to help you design new materials and processes
 - to understand the outcomes of real experiments
- In the best sense of reproducible research
 - By keeping software versions and input data with the results
 - Including the post-processing options used
- While making a living for ourselves
- If you find you'd like to know more after this presentation, come to our booth in the exhibition

WHAT WOULD WE LIKE TO ACHIEVE?

GEODICT



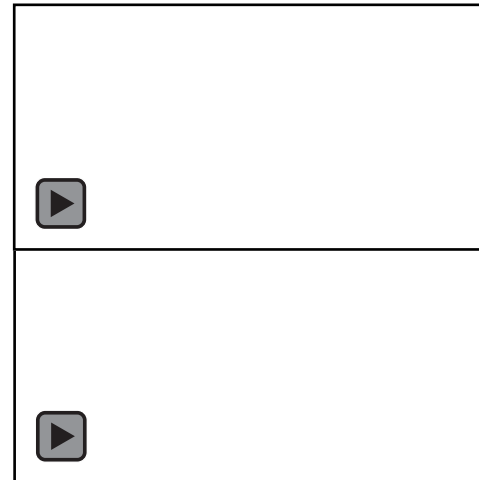


- Obtain grammage total, per material or per class of objects
- Output of objects as triangulation (.stl), CAD (.dxf), etc.
- Find number of contact points per objects
- Determine shape & dimensions of individual objects
- Compute object orientation distributions
- Etc. etc.

HOW DO WE LABEL CT SCANS?

- *Now*: segment 3D scan; then label it using Artificial Intelligence
- **Soon**: label scan (gray values) directly with Artificial Intelligence

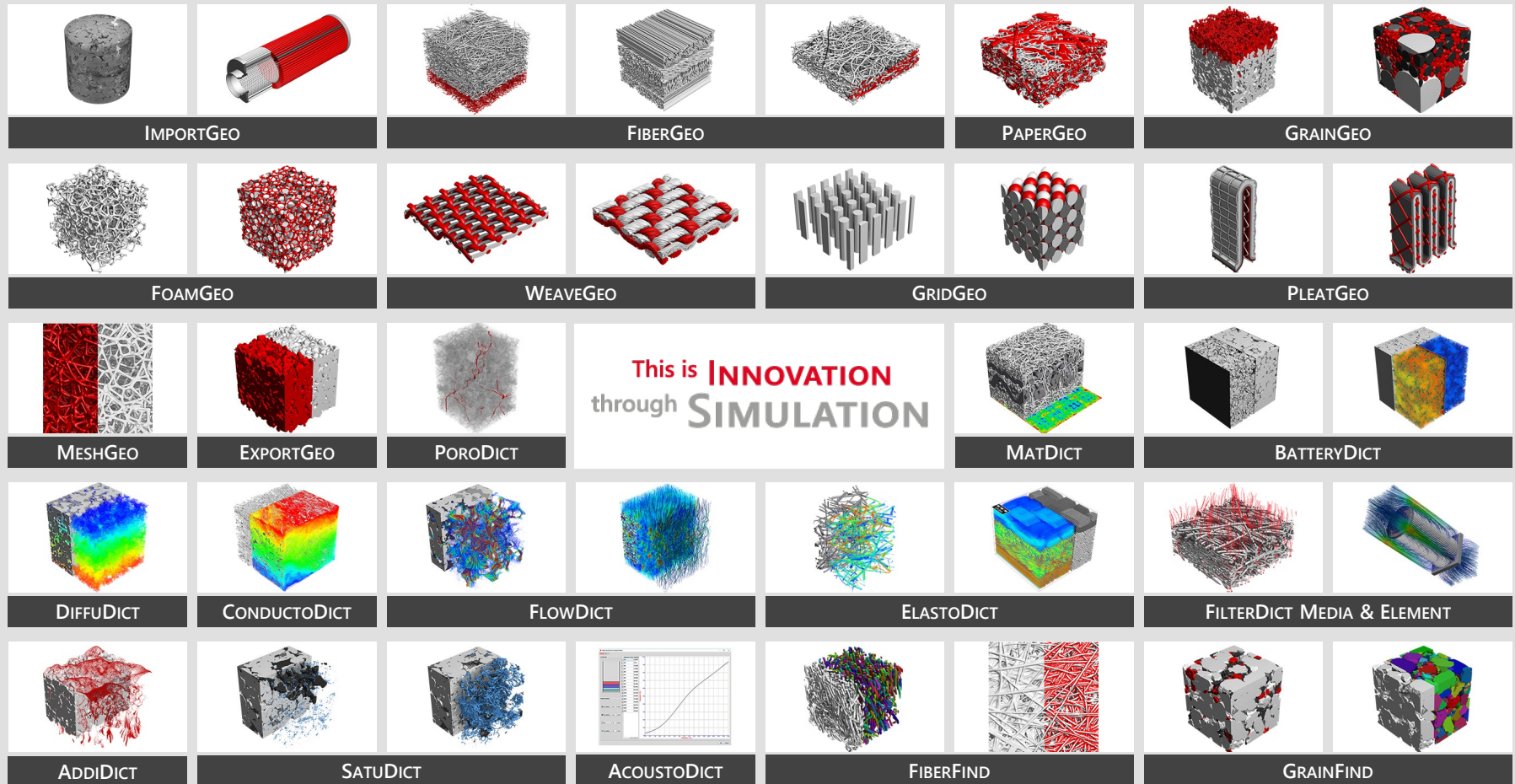
- *Now*: segment 3D scan; then label it using Artificial Intelligence
- **Soon**: label scan (gray values) directly with Artificial Intelligence
- AI requires training data / ground truth
- Segmented training data can be created with GeoDict today
- Gray value images can be created with methods from Monday's presentation by Andreas Weber
- Those will be use to train the next generation neural networks in GeoDict

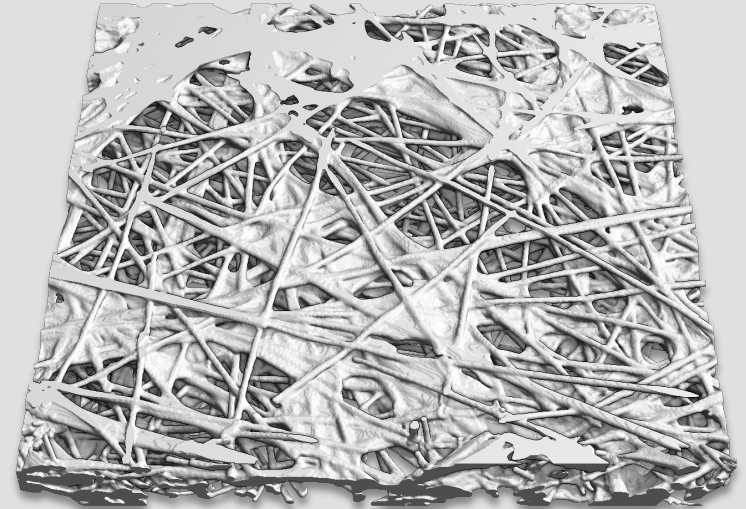
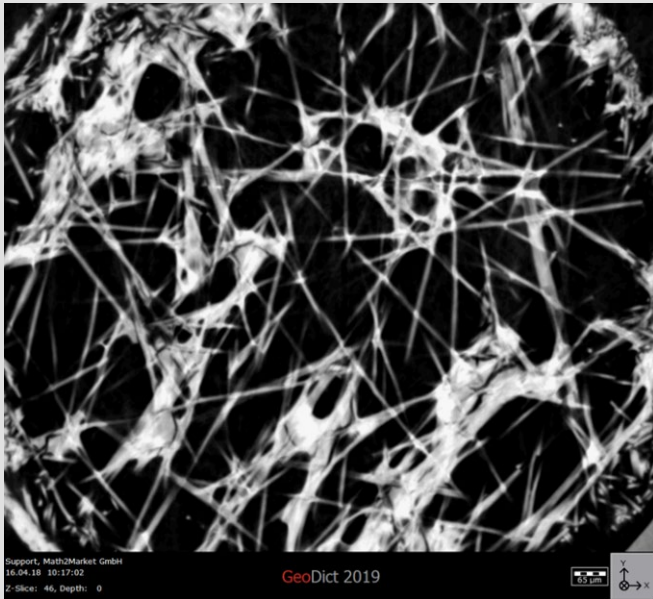


GEODict® MODULE OVERVIEW

A RICH SOURCE FOR AI TRAINING DATA

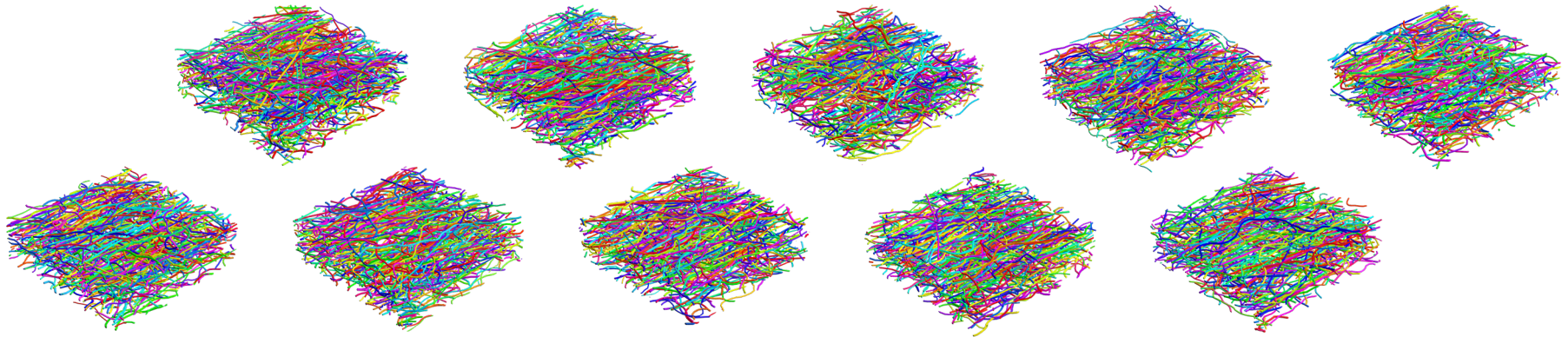
GEODict





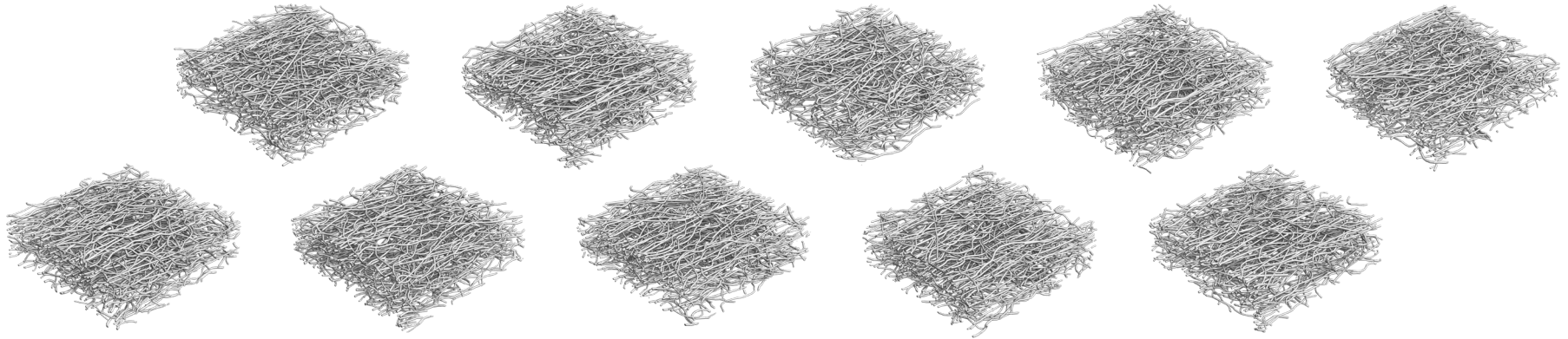
From stack of gray value slices

To 3-D empty / solid image



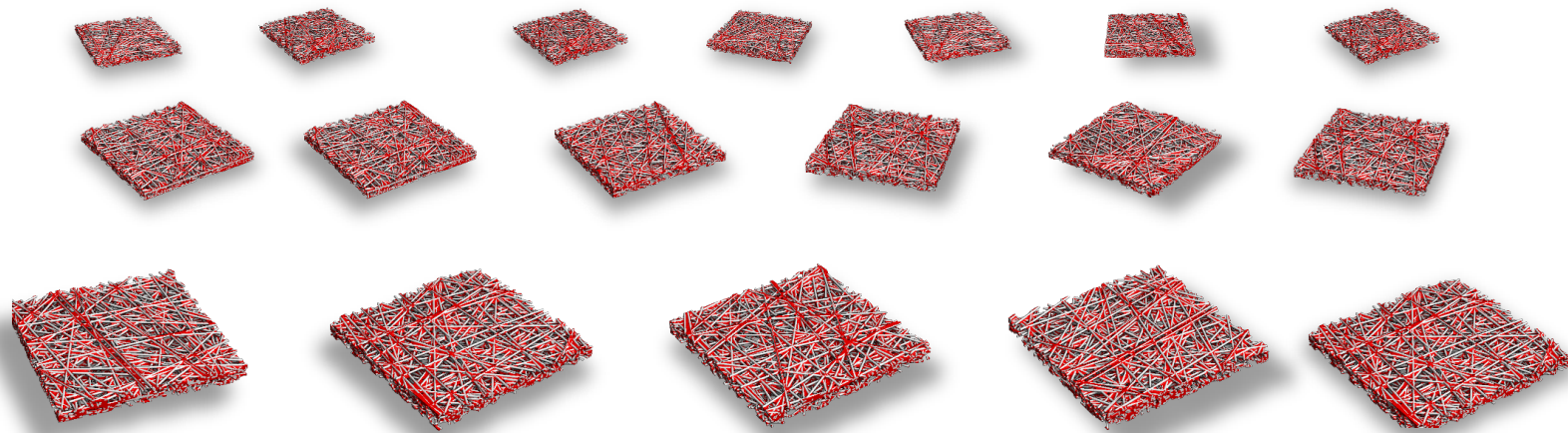
Training Data: Use GeoDict's unique fiber modelling capabilities:

- Modeled 10 Digital siblings (512x512x256 Voxels) as training data
- Varied fiber curvature, orientation, length and diameter
- Corresponded to ~1 billion solid voxels as training data points



Training Data: Then make the models look like binarized scans!

- All fibers in the models get the same gray value, just as in the segmented 3D scans

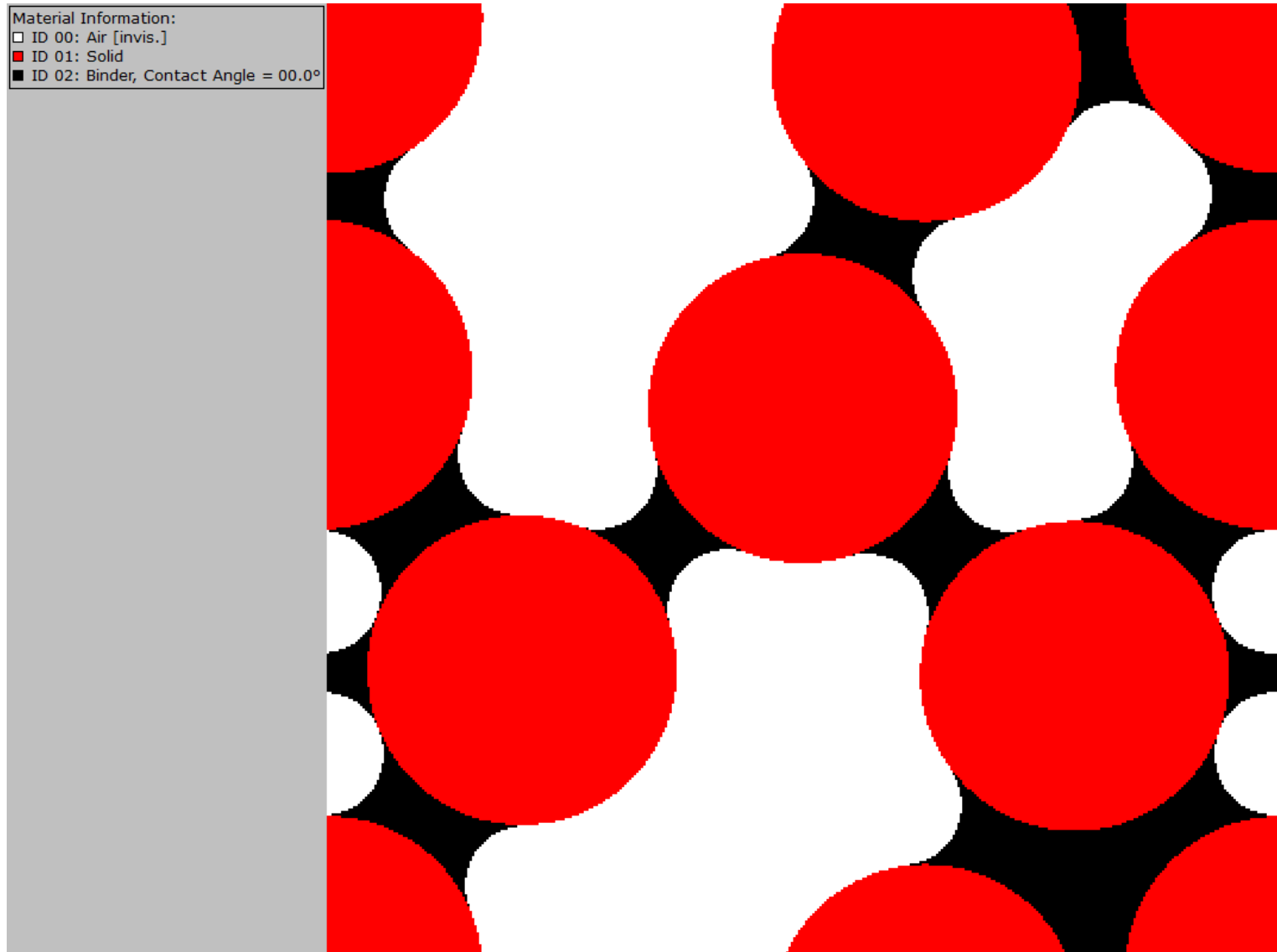


Solution: Use GeoDict's unique material modelling capabilities

- modeled 18 Digital siblings (512x512x256 Voxels) as training data
- varied porosity, binder parameters as estimated for 4 different (Toray GDL) samples
- corresponds to ~800 million solid voxels as training data points

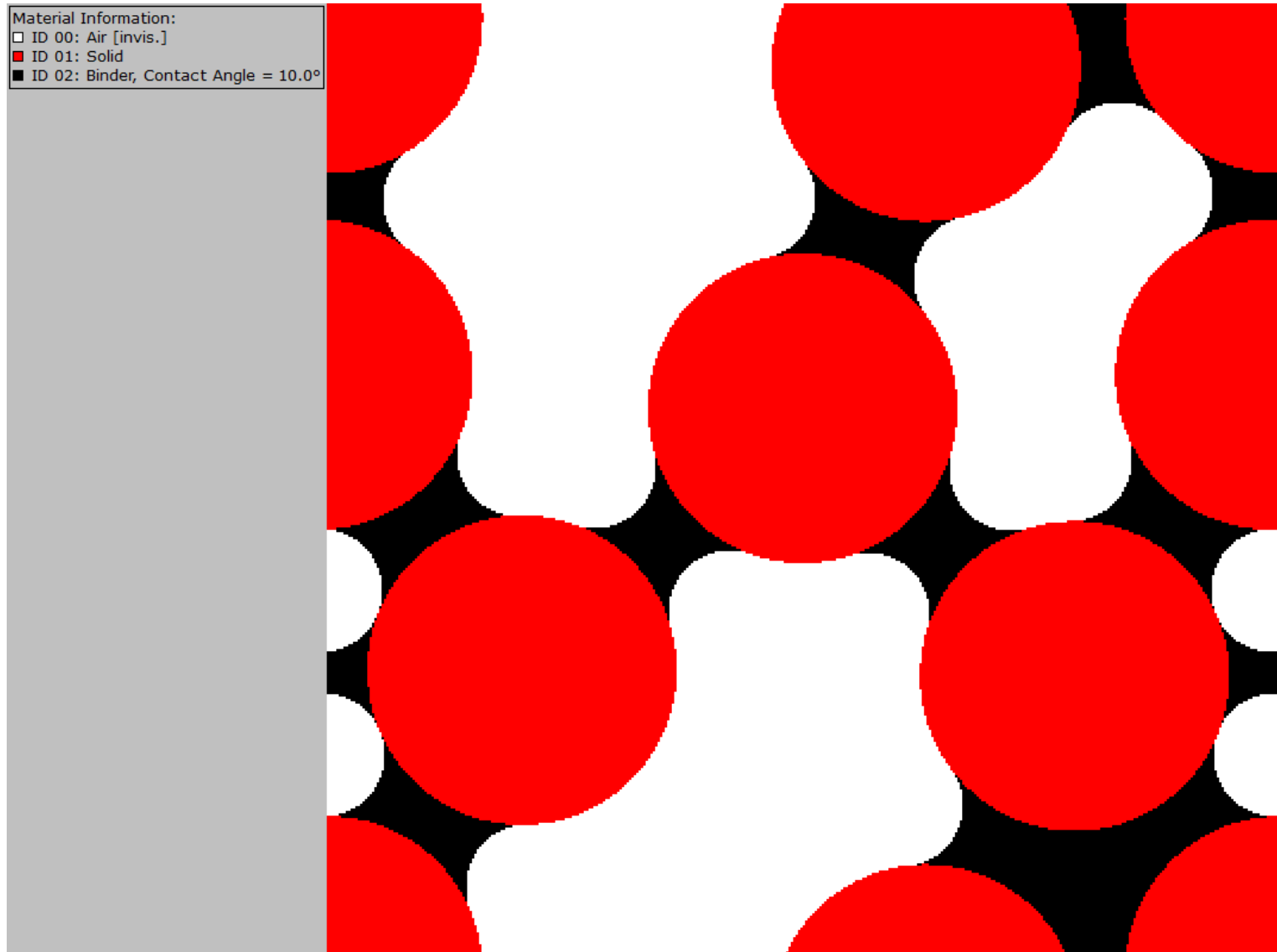
ADDING BINDER: CONTACT ANGLE 0°

GEODICT



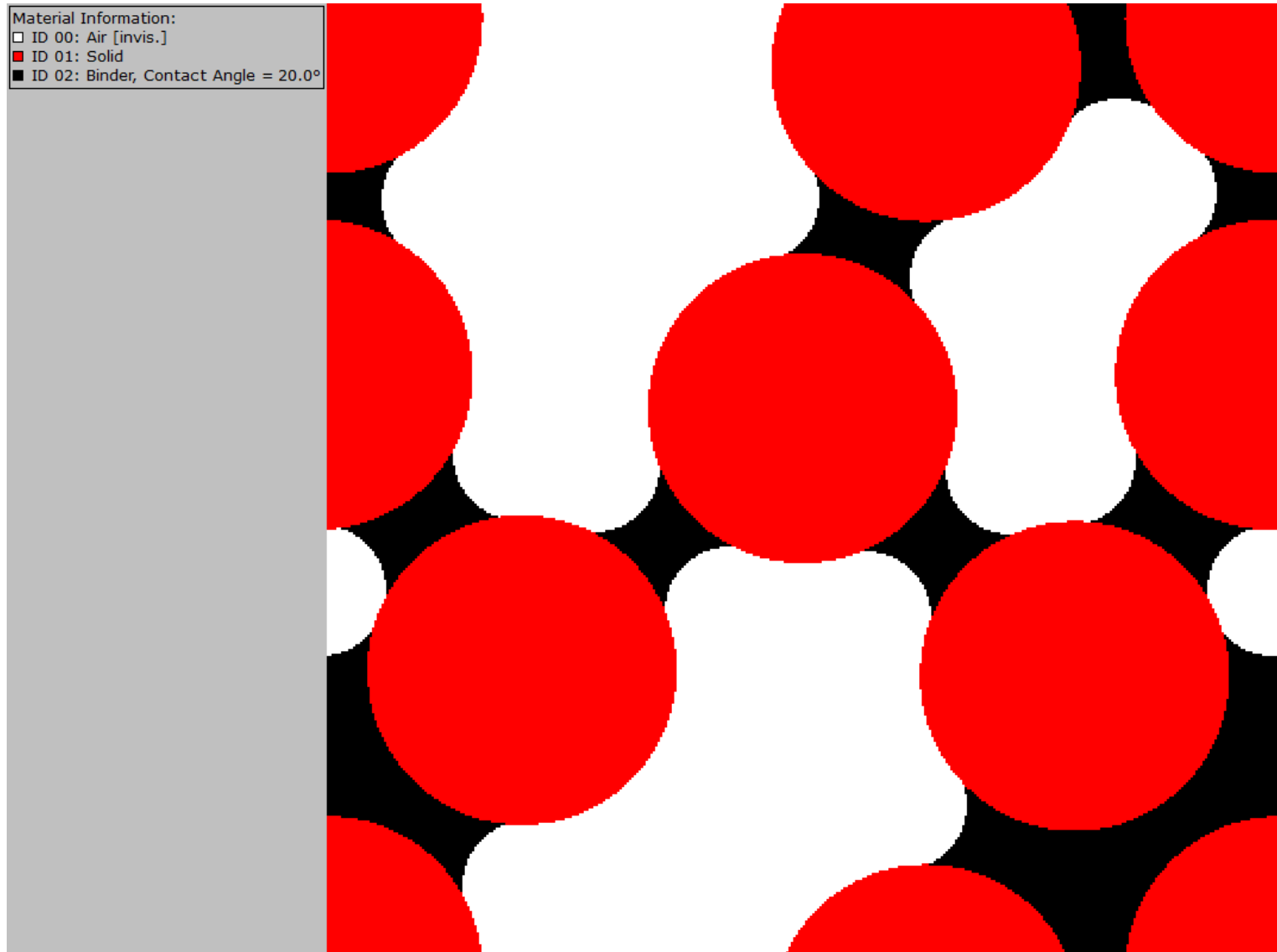
ADDING BINDER: CONTACT ANGLE 10°

GEODICT



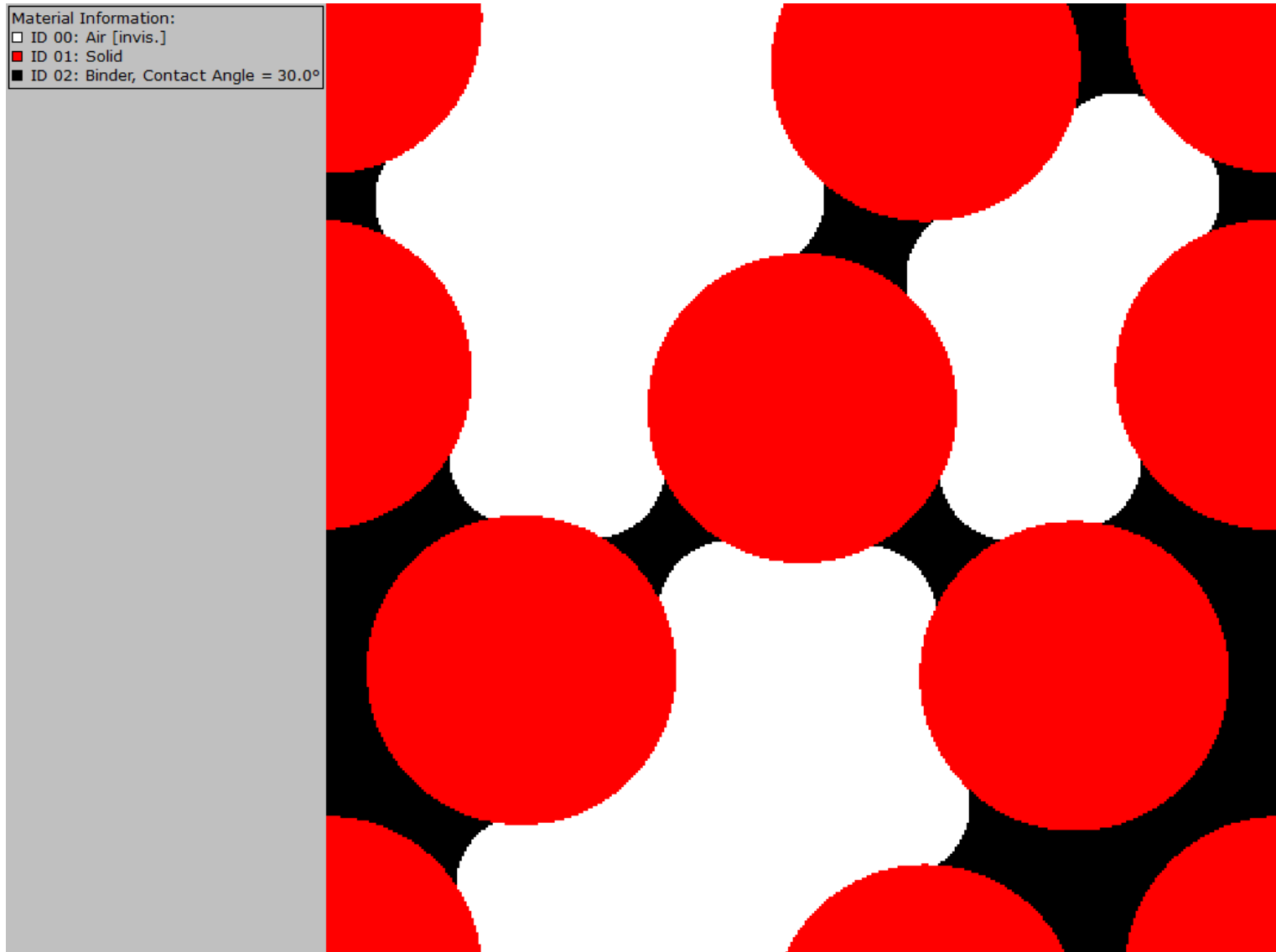
ADDING BINDER: CONTACT ANGLE 20°

GEODICT



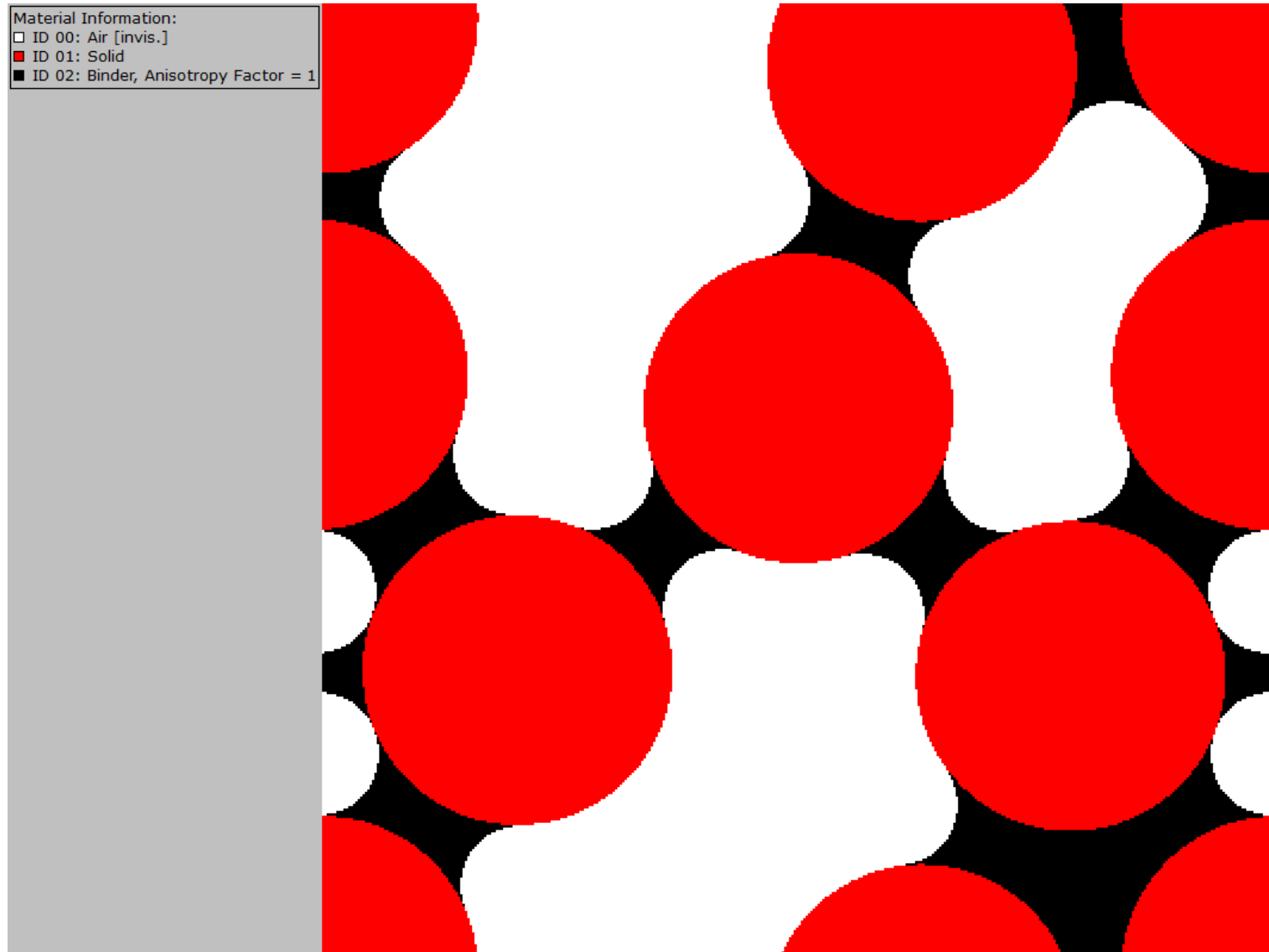
ADDING BINDER: CONTACT ANGLE 30°

GEODICT



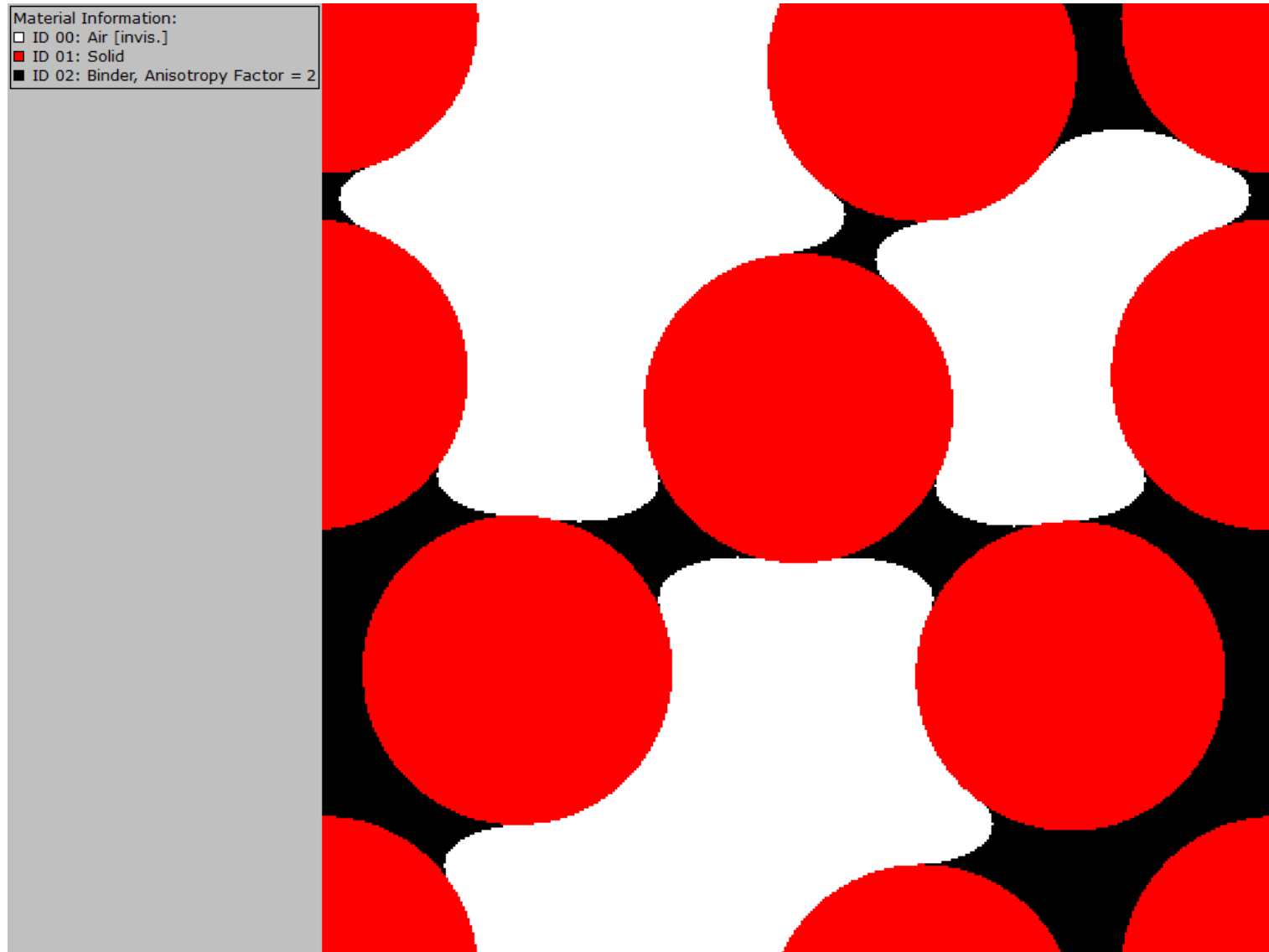
ISOTROPIC BINDER (ANISOTROPY FACTOR 1)

GEODICT



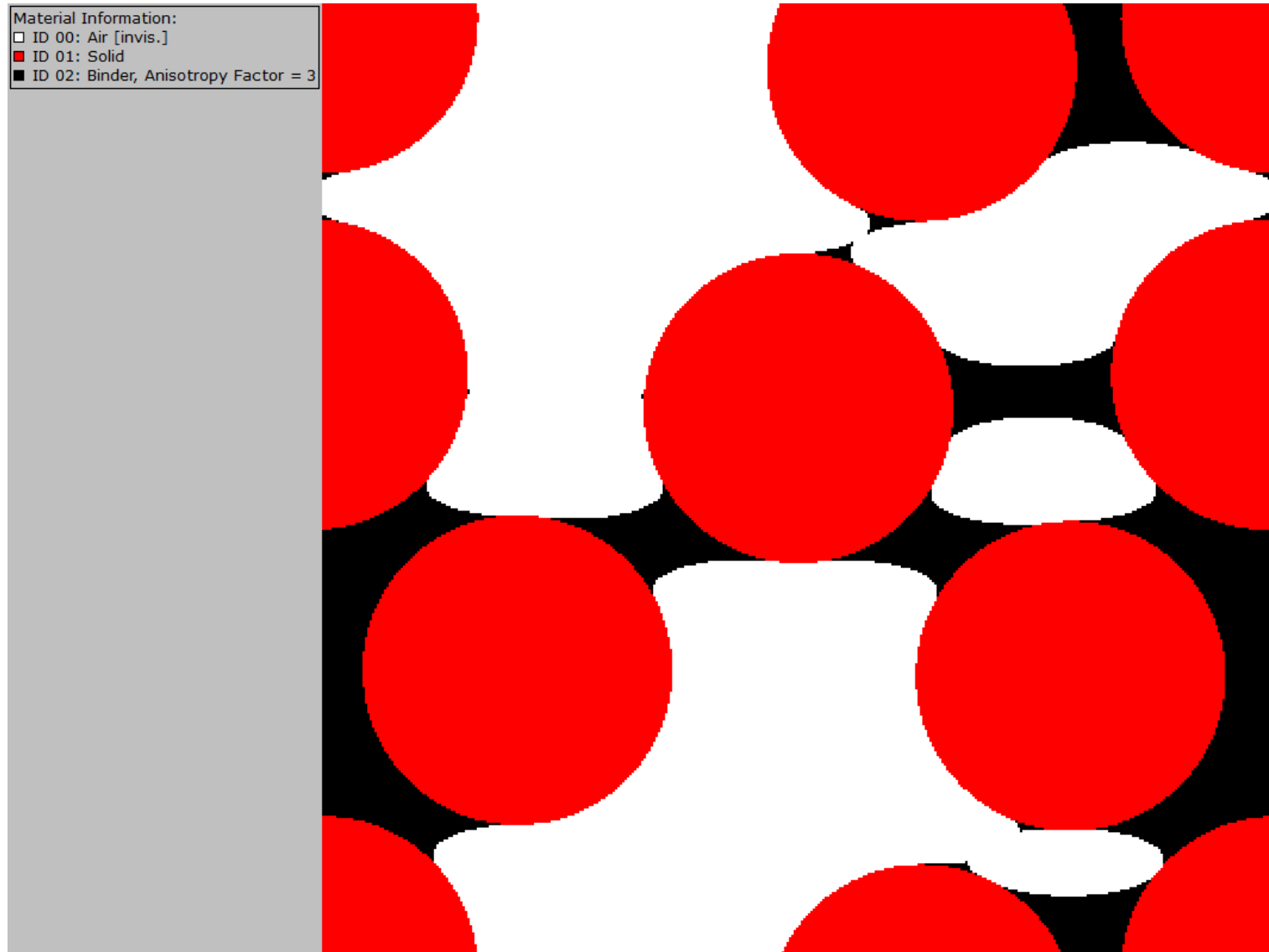
ANSOTROPIC BINDER (ANISOTROPY FACTOR 2)

GEODICT



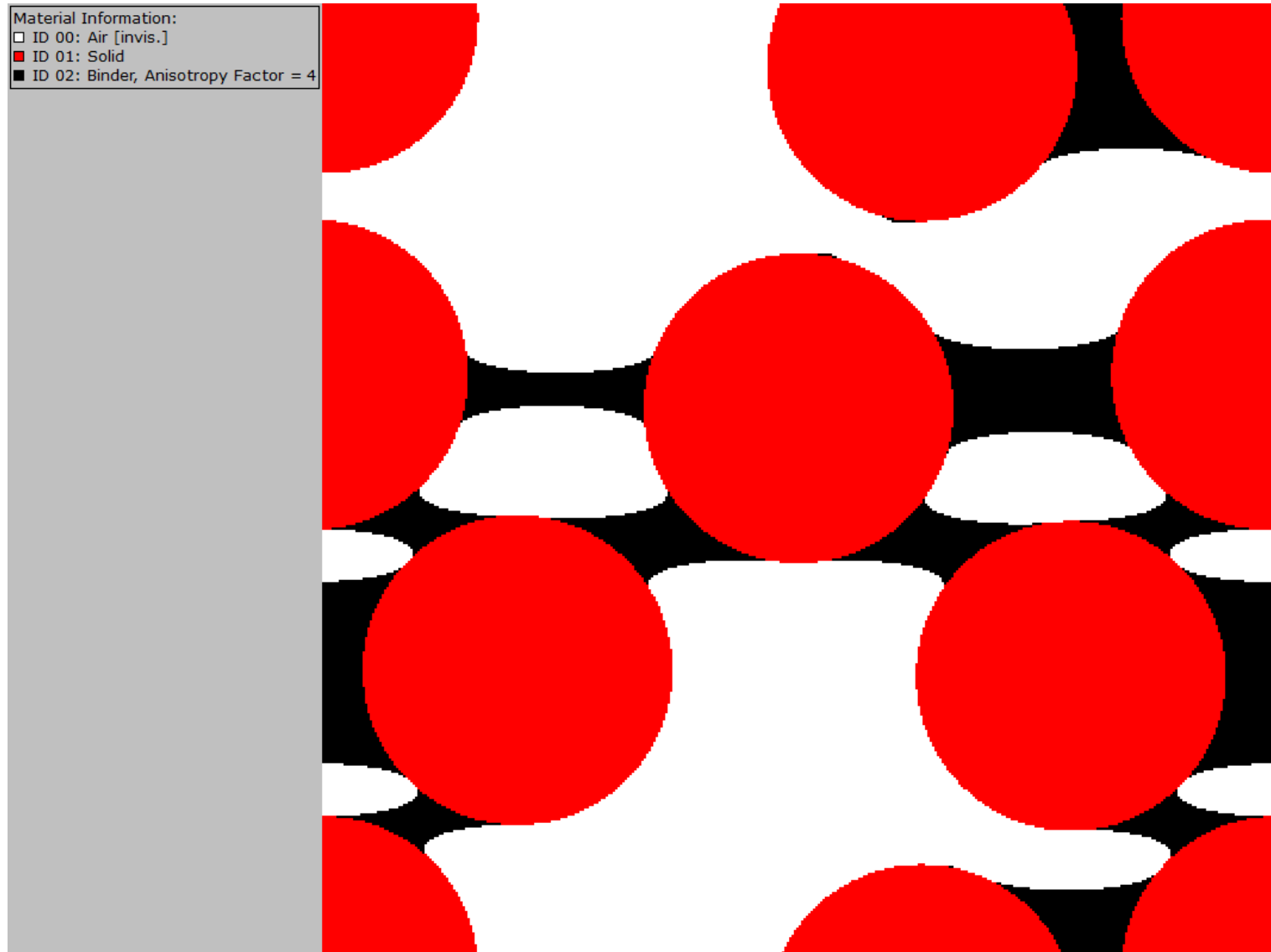
ANSOTROPIC BINDER (ANISOTROPY FACTOR 3)

GEODICT

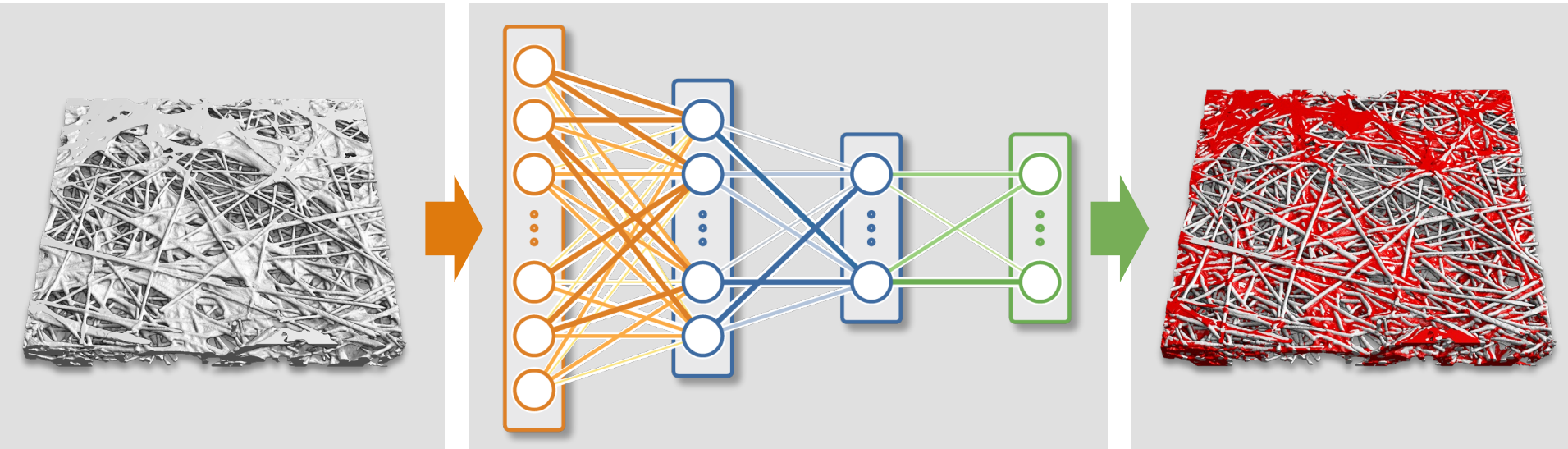


ANSOTROPIC BINDER (ANISOTROPY FACTOR 4)

GEODICT



NEURAL NETWORK (N.N.) TRAINING & USAGE PHASES



Training: N.N. learns edge weights from input and output

- input: Digital Twin data: binarized version
- output: Digital Twin data: labeled binder and fibers version

Usage: N.N. predicts labeled output from input and edge weights

- input: 3D scan data: binarized
- output: 3D scan data: labeled binder and fibers

TORAY PAPER TGP-H-030, 05% WET PROOFING

GEODICT



TORAY PAPER TGP-H-030, 10% WET PROOFING

GEODICT



TORAY PAPER TGP-H-030, 30% WET PROOFING

GEODICT



TORAY PAPER TGP-H-030, 50% WET PROOFING

GEODICT



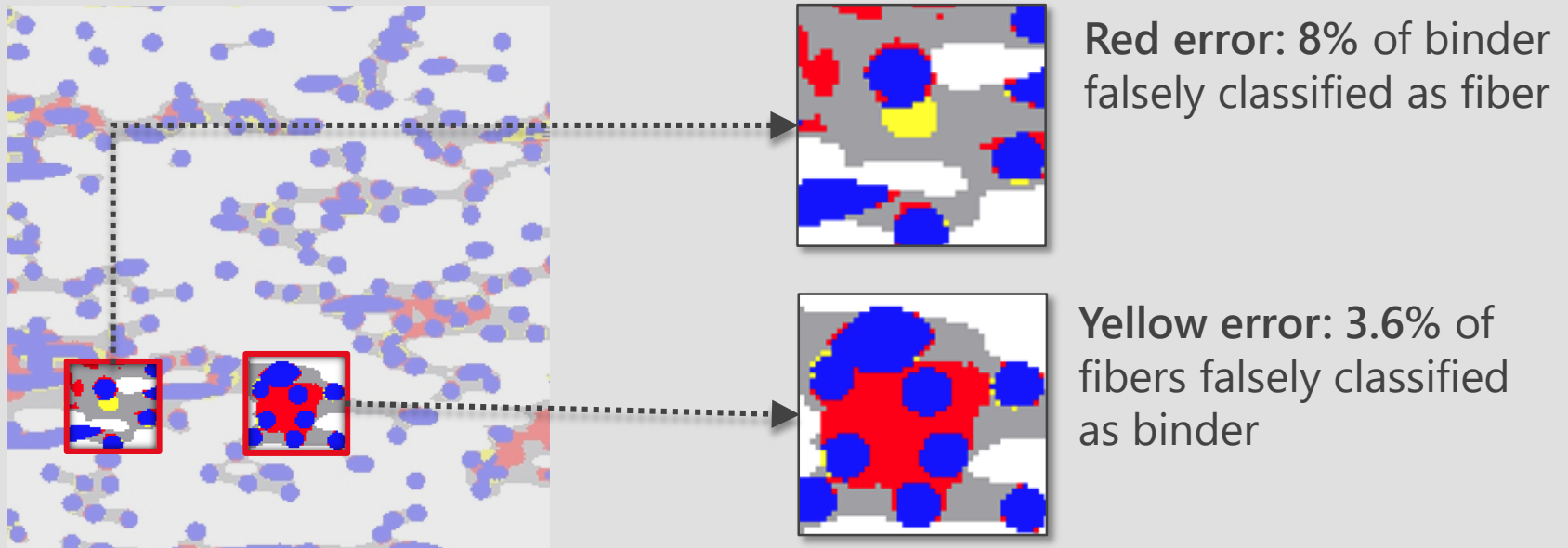
BINDER IDENTIFICATION IN GAS DIFFUSION LAYER

Crossection in X-Direction:



Idea: Maybe create 3-D Digital Twins from 2-D images using AI in the future?

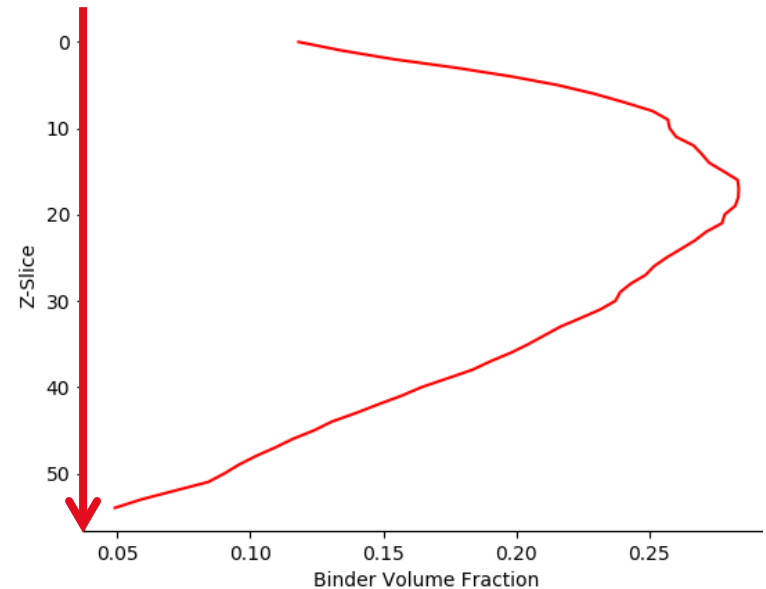
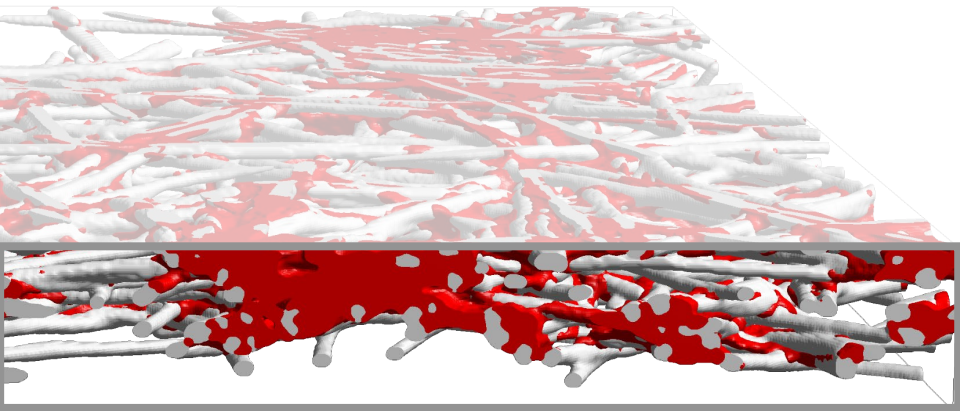
VALIDATION – MISSCLASSIFICATION ANALYSIS OF BINDER / FIBERS



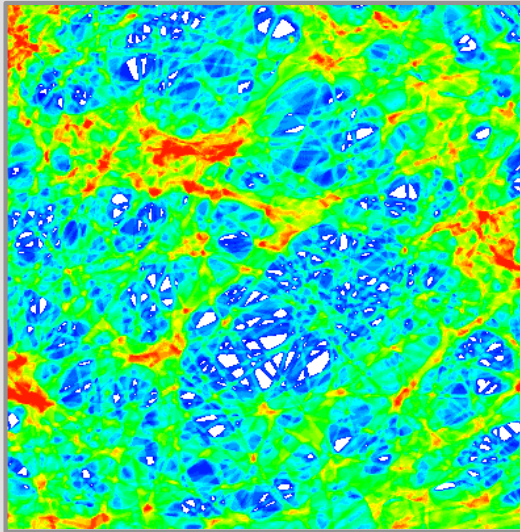
Direct comparison of ground truth Digital Twin to n.n. result

- Blue/gray/white: correctly identified fiber/binder/pore
- Total binder volume percentages: True 13.1%, predicted 11.5%

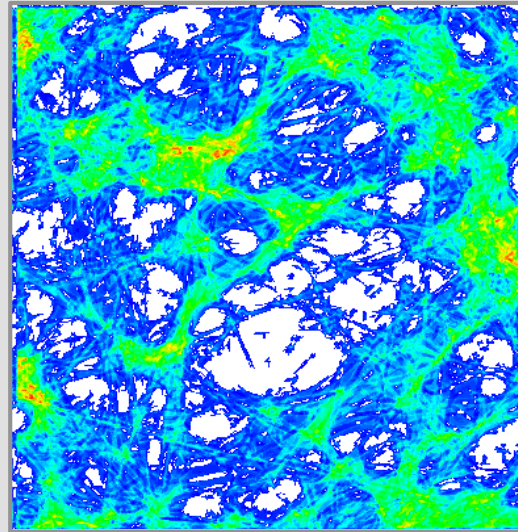
- Grammage / volume percentages: total, binder only or fibers only
- Binder distribution in through-plane and in-plane directions
- Size-distribution of binder points
- Total number of binder blobs/components
- Output of binder or fibers as triangulation (.stl), CAD (.dxf), etc.
- Number of contact points per fiber
- Length of individual fibers
- Curvature distribution along individual fibers



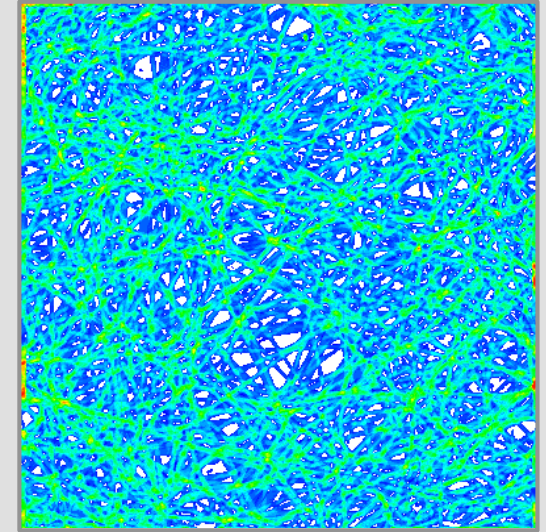
- In production, binder is applied to the top of the fiber and, then, intrudes into deeper layers
- After labelling binder voxels, we can compute the distribution of binder in through direction (right)



Both materials



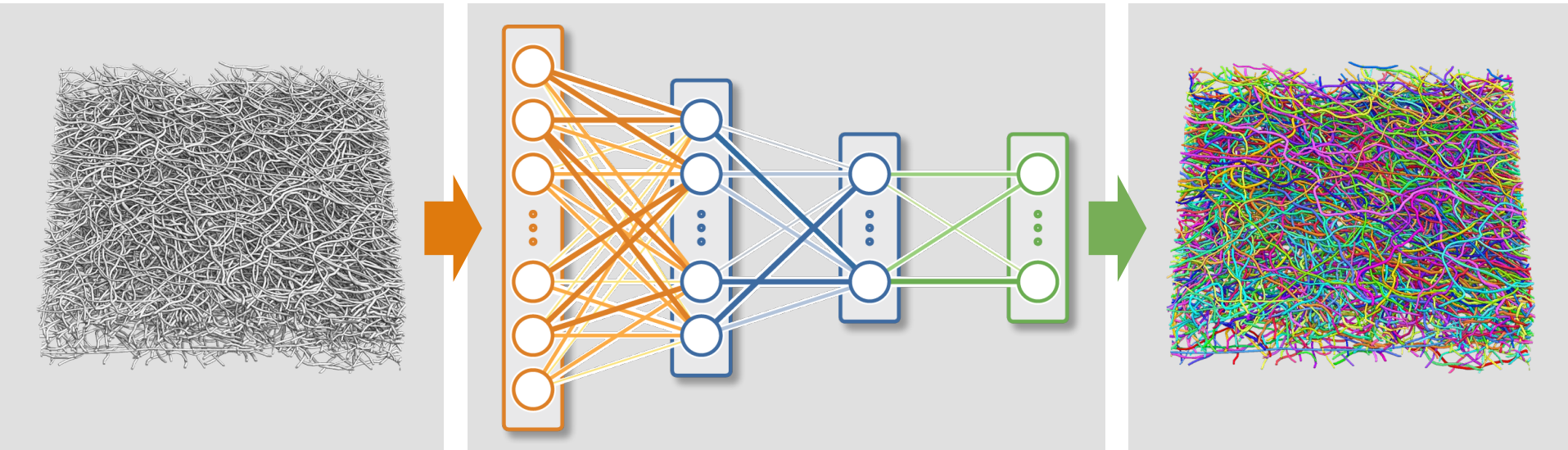
Binder only



Fibers only

Scale: White: 0% volume fraction, Red: 7.2% volume fraction

NEURAL NETWORK (N.N.) TRAINING & USAGE PHASES

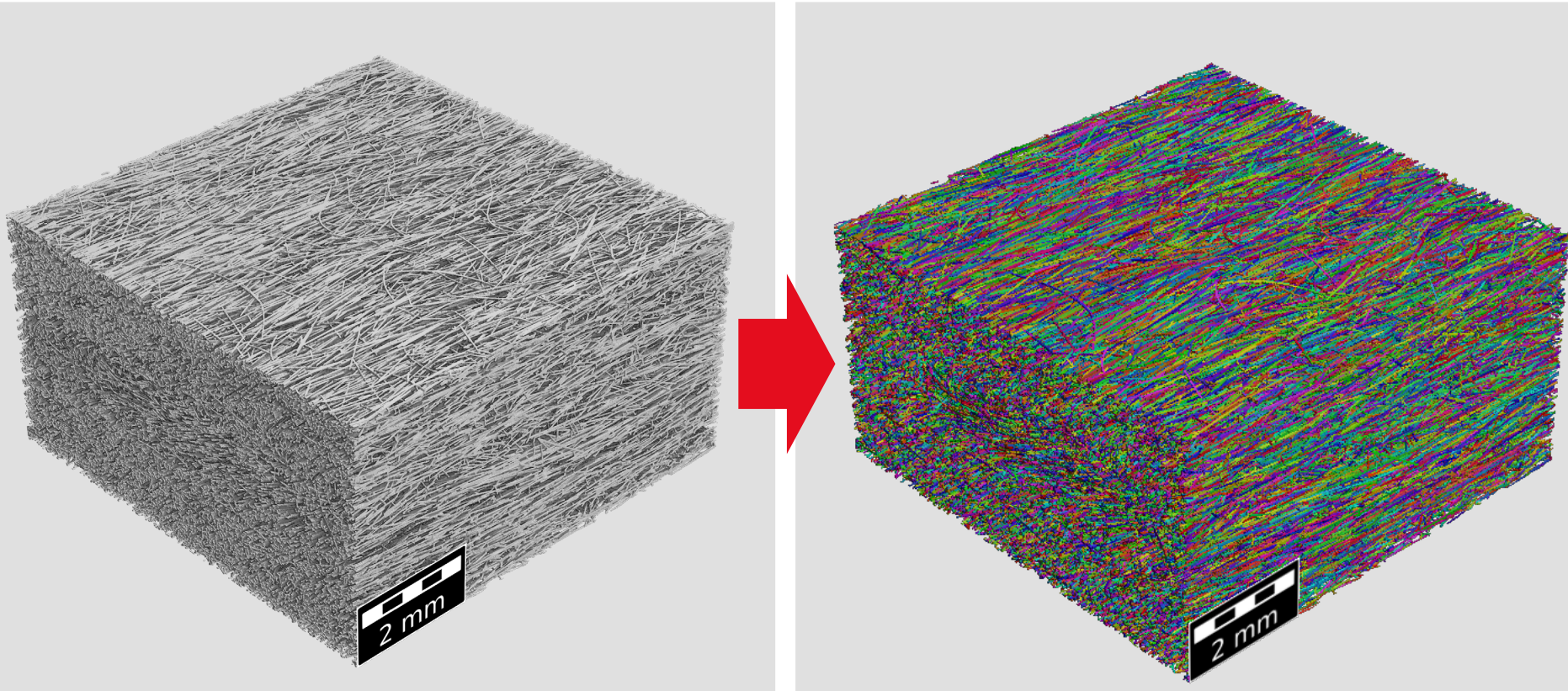


Training: N.N. learns edge weights from input and output

- input: Digital Twin data: binarized version
- output: Digital Twin data: labeled fibers version

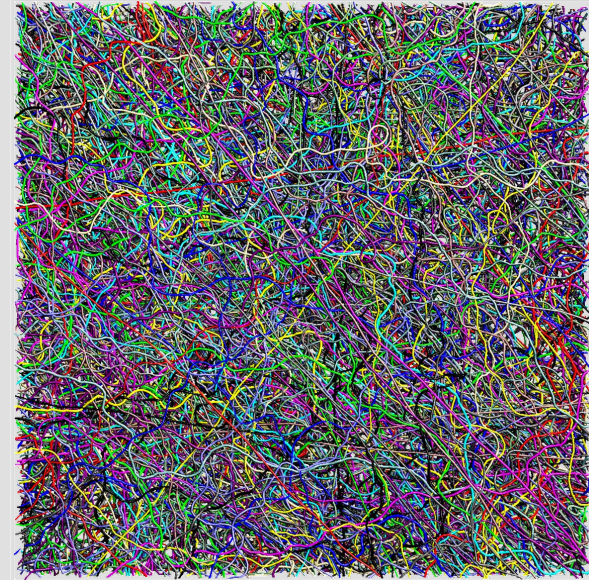
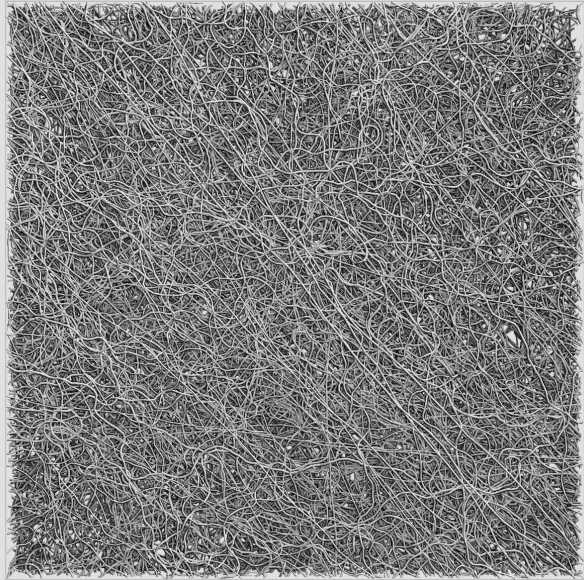
Usage: N.N. predicts labeled output from input using weights

- input: 3D scan data: binarized
- output: 3D scan data: labeled fibers



Input: segmented μ CT scan

Output: labeled fibers



Input: segmented μ CT-Scan

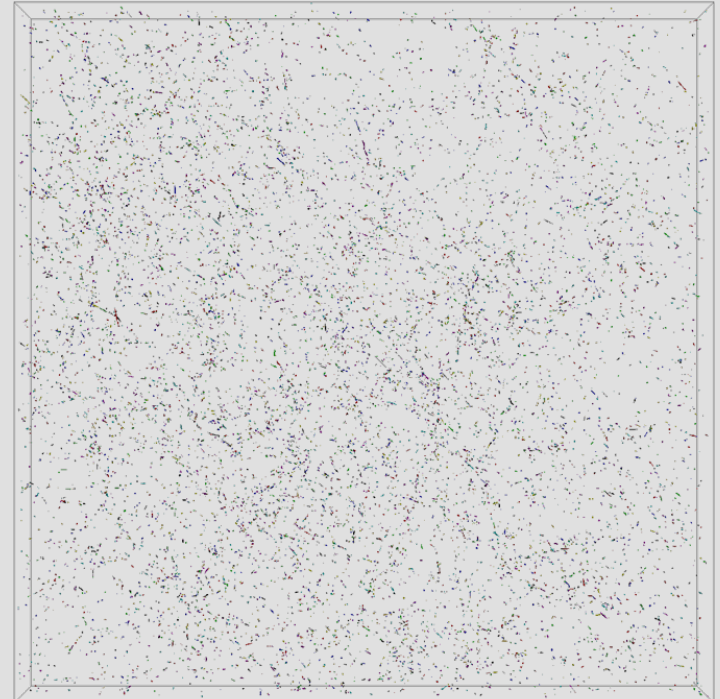
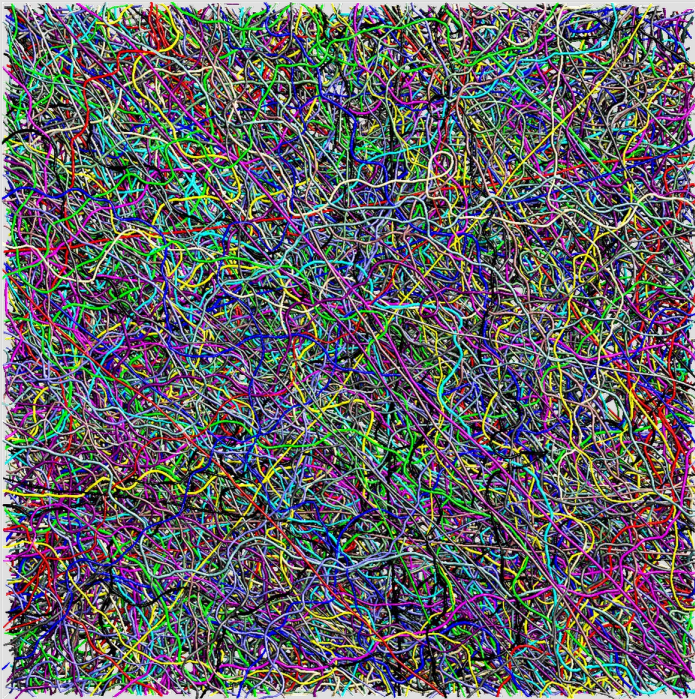
Output: labeled fibers

MORE EXAMPLES OF COMPUTABLE QUANTITIES:

- Grammage / volume percentages:
total, binder separate or fibers separate
- Binder distribution in through-plane and in-plane directions
- Size-distribution of binder points
- Total number of binder blobs/components
- Output of binder or fibers as triangulation (.stl), CAD (.dxf), etc.
- Number of contact points per fiber
- Length of individual fibers
- Curvature distribution along individual fibers
- Fiber orientation distribution

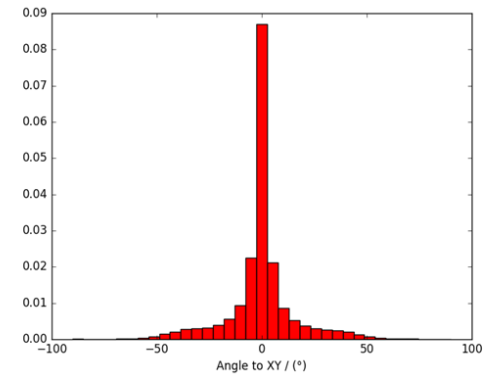
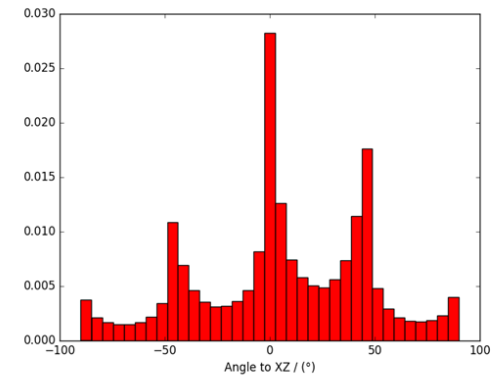
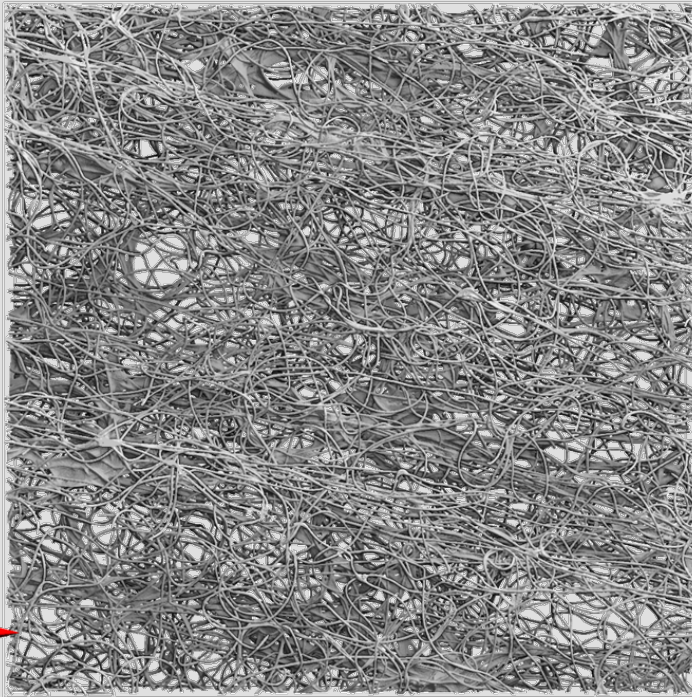
DETECTION OF 14995 BOND POIN

GEODICT



Input: labeled individual fibers

Output: labeled individual
bonds



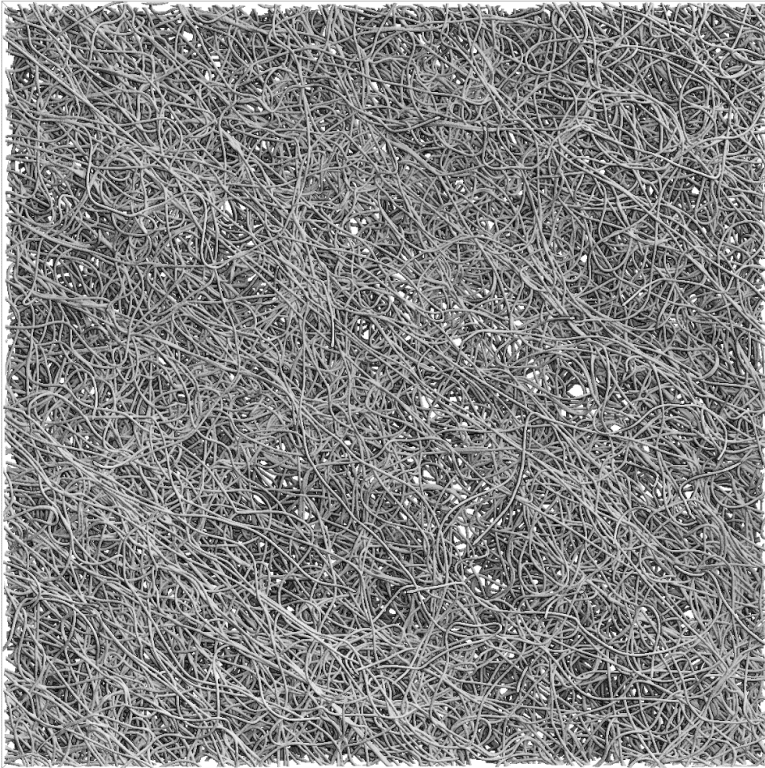
Input: segmented CT scan

Output: orientation plots

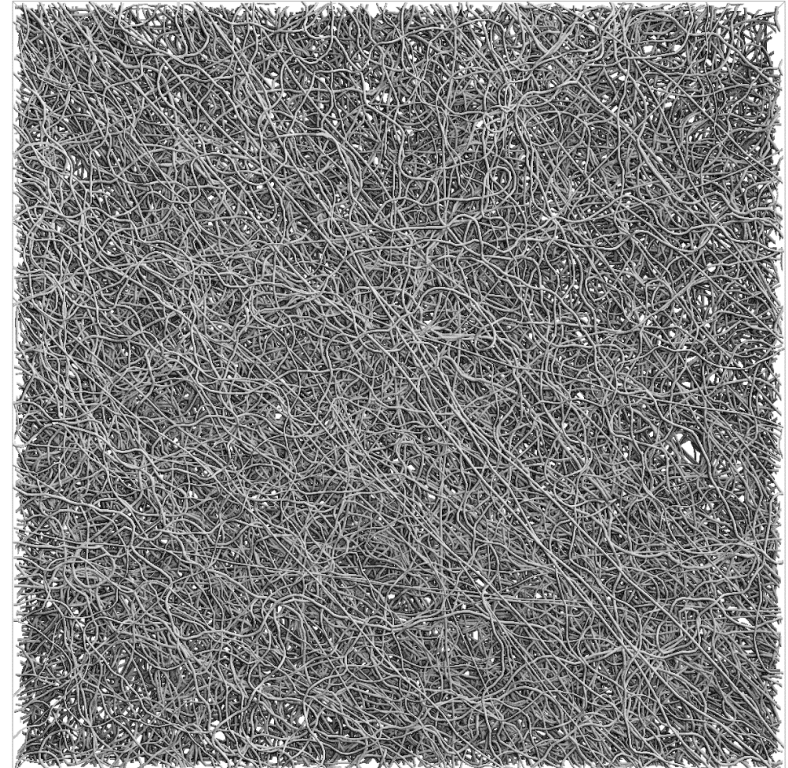
- Next three slides show manually created statistical (digital) twins
- This is right now one of the bottle necks: n.n. need models close to the real data
- One of the future works is to use generative adversary networks g.a.n. to create the models
- Another future work is the creation of gray value images from models that include the typical artefacts like rings and noise to use as left side in the training and usage phases of **FiberFind**-AI.

CT-SCAN VS DIGITAL TWIN GENERATED IN GEODICT

GEODICT



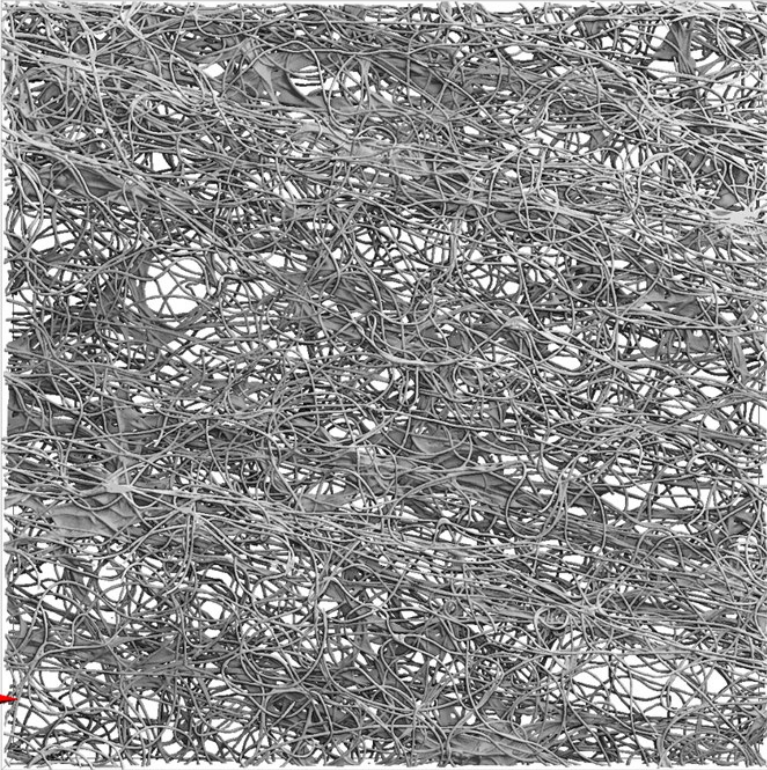
μCT-scan



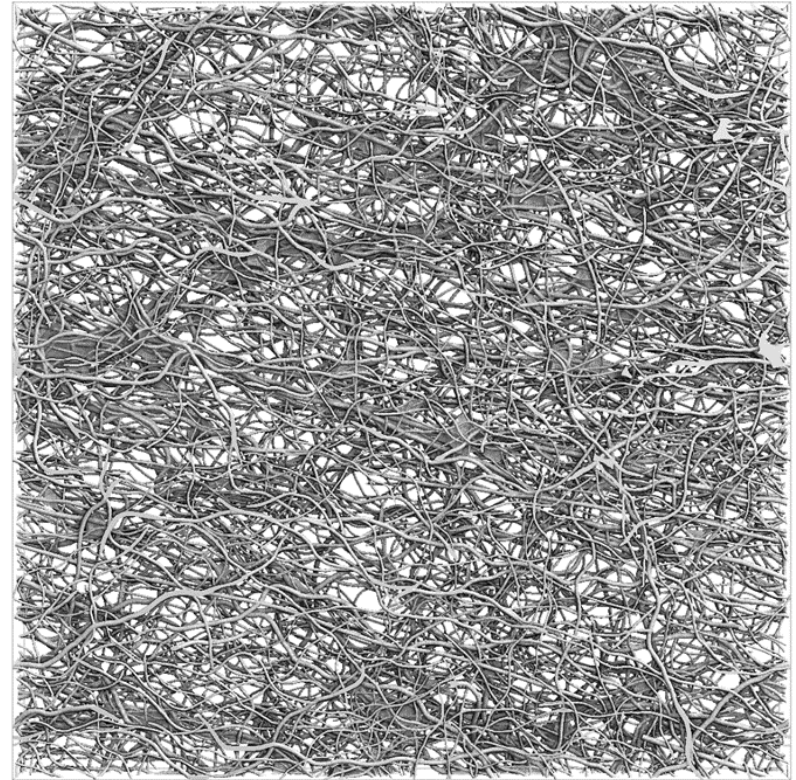
Digital Twin

CT-SCAN VS DIGITAL TWIN GENERATED IN GEODICT

GEODICT



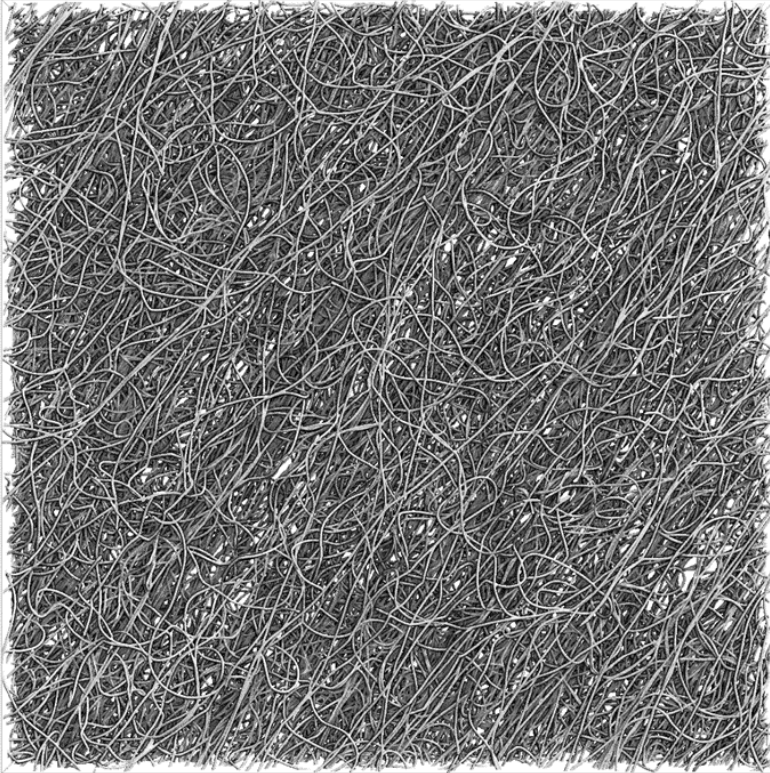
μCT-scan



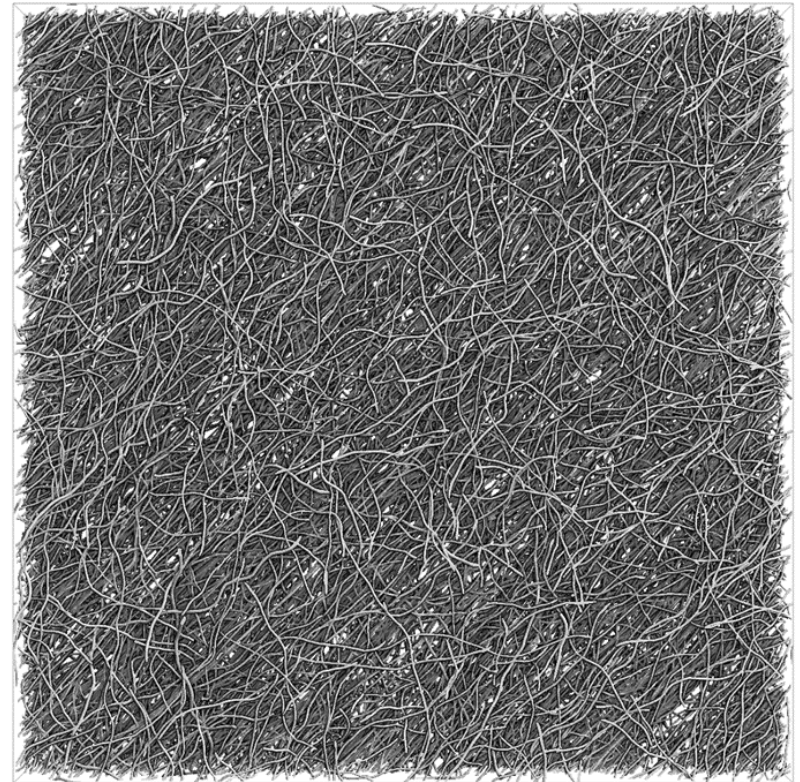
Digital Twin

CT-SCAN VS DIGITAL TWIN GENERATED IN GEODICT

GEODICT



μCT-scan

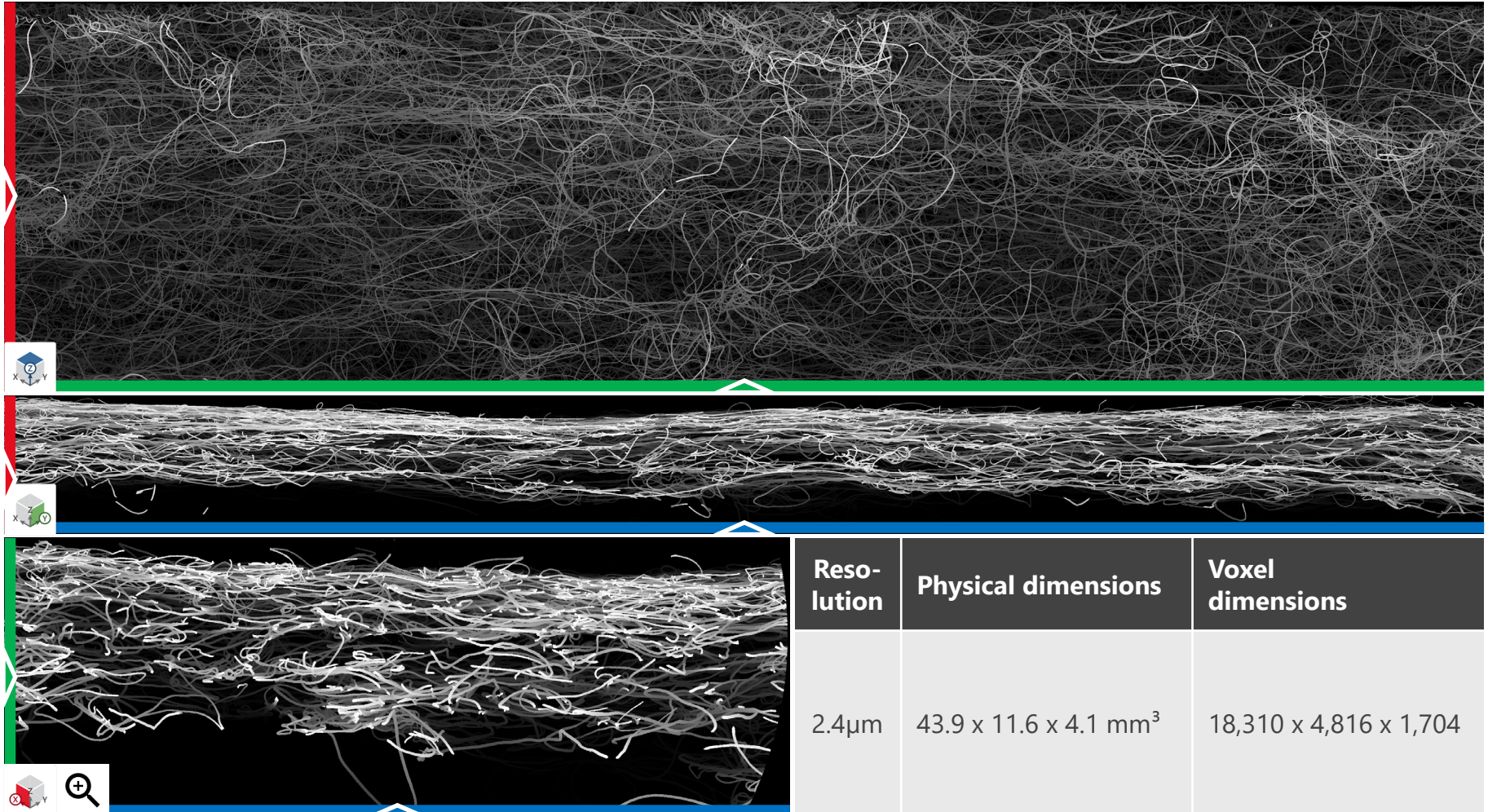


Digital Twin

Sample Name	Resolution	Physical dimensions	Domain sizes in voxels
A	2.4μm	43.9 x 11.6 x 4.1 mm	18,310 x 4,816 x 1,704 = 150 Giga Vox
B	2.7μm	42.2 x 10.9 x 4.8 mm	15,619 x 4,032 x 1,796 = 113 Giga Vox

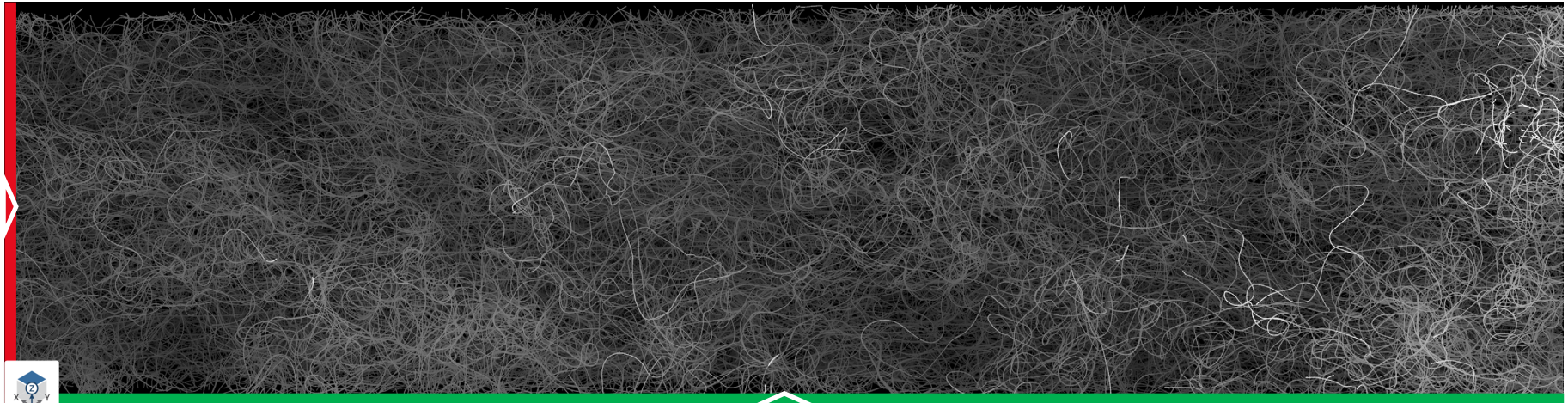
- Carded nonwoven samples
- Scanned and stitched together by Bruker microCT
- Analyzed by Math2Market on workstation with 1TB of memory

SAMPLE A – SEM VIEW



Resolution	Physical dimensions	Voxel dimensions
2.4μm	43.9 x 11.6 x 4.1 mm ³	18,310 x 4,816 x 1,704

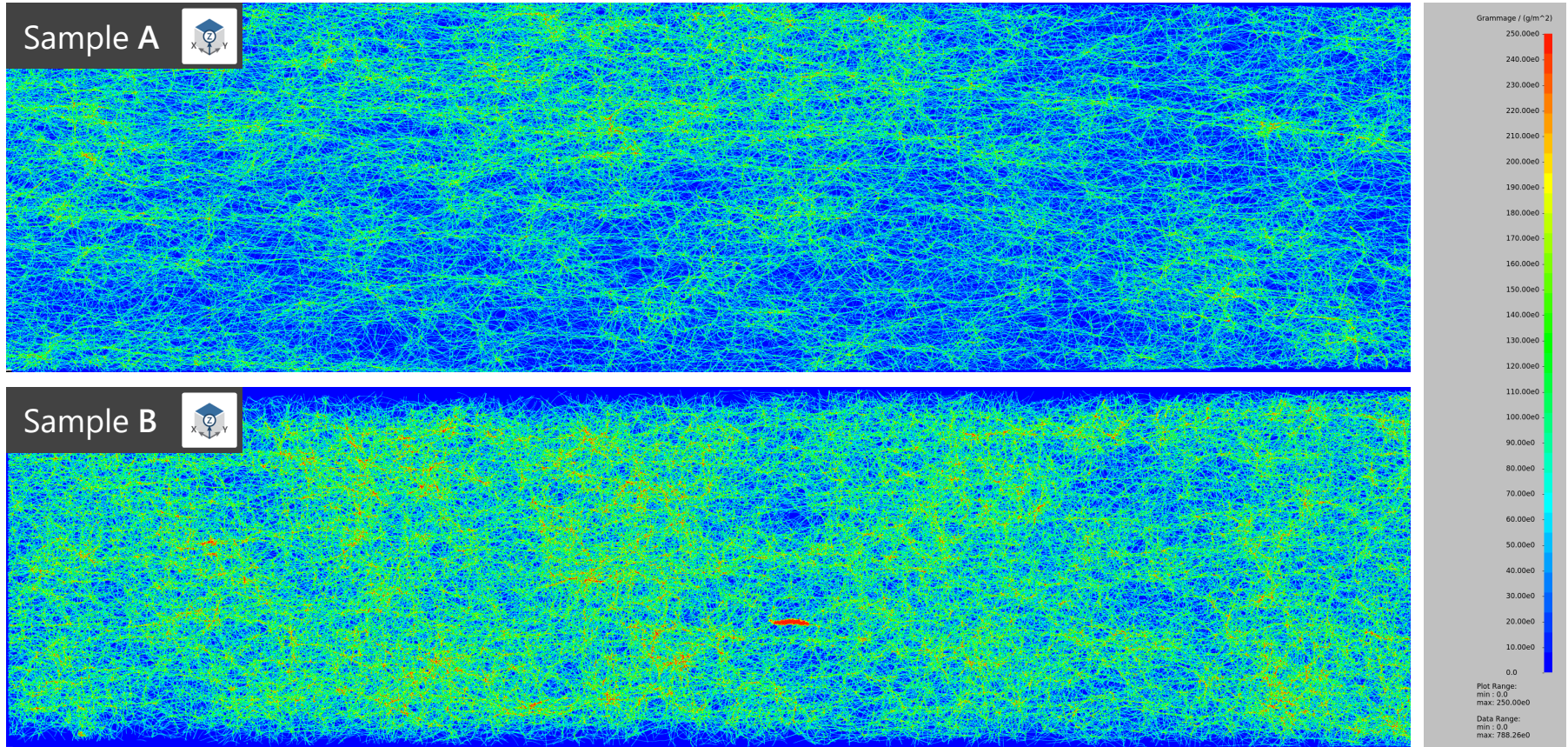
SAMPLE B – SEM VIEW



Resolution	Physical dimensions	Voxel dimensions
2.7 μ m	42.2 x 10.9 x 4.8 mm ³	15,619 x 4,032 x 1,796



DENSITY MAP (CLOUDINESS)



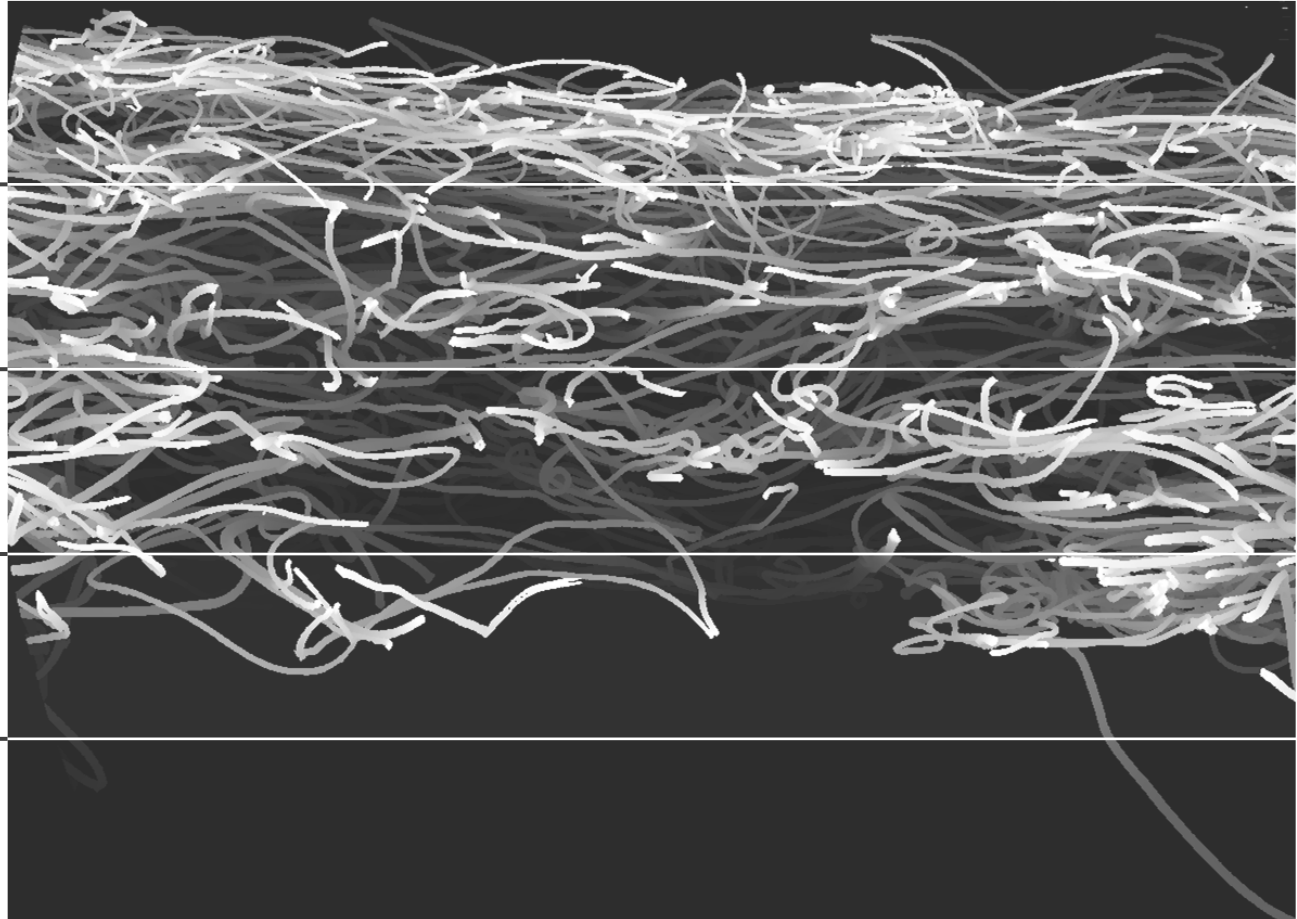
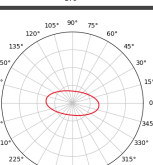
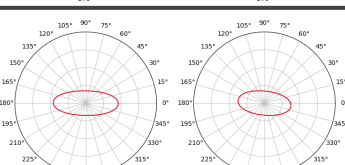
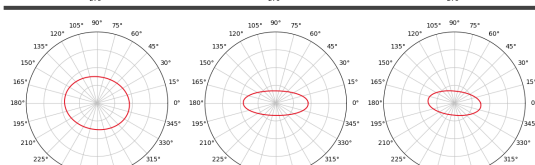
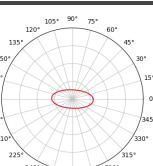
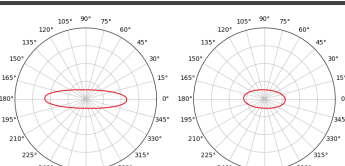
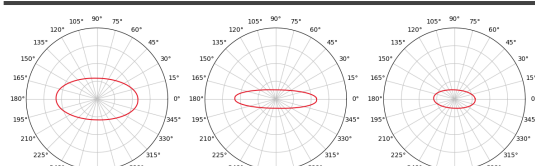
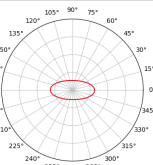
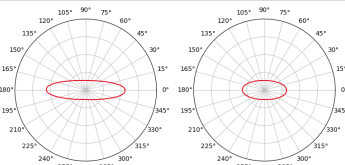
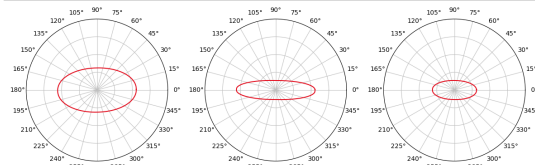
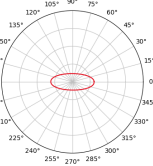
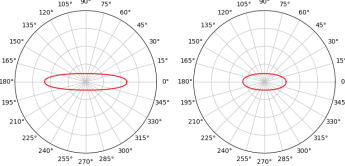
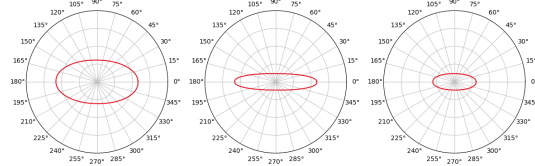
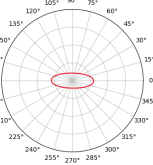
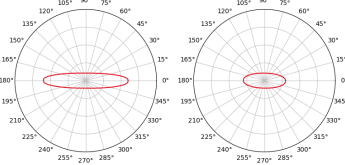
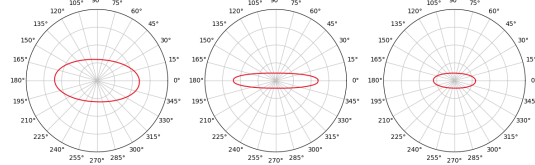
FIBER ORIENTATIONS – SAMPLE A

XY

XZ

YZ

View of a section of the surface in the direction of the X axis.



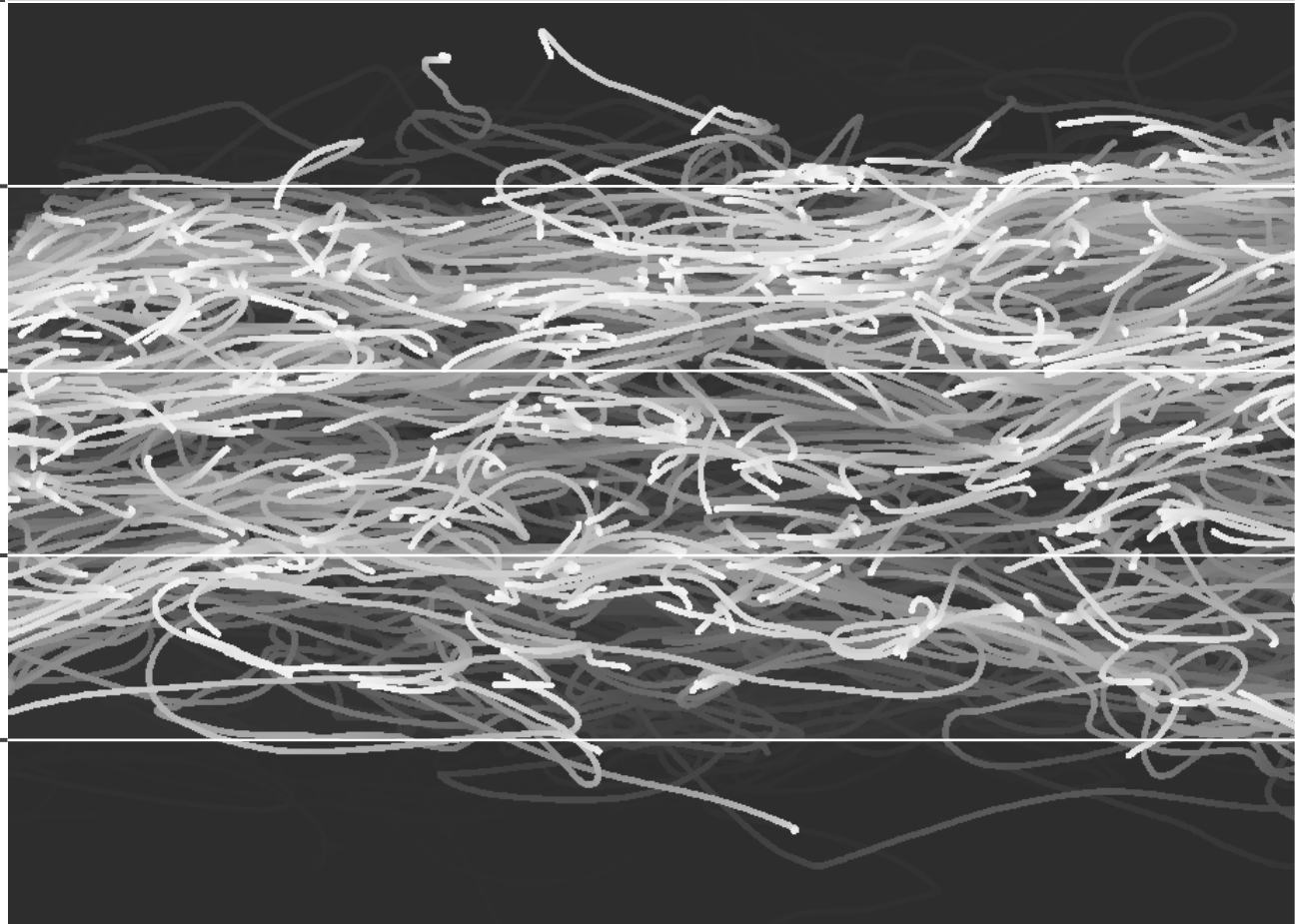
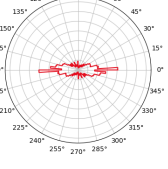
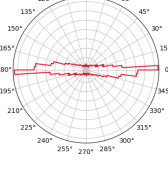
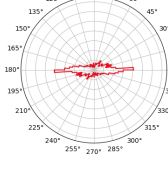
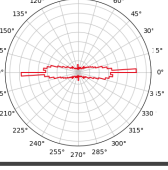
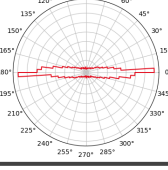
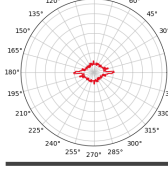
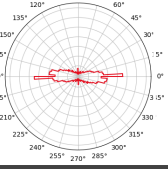
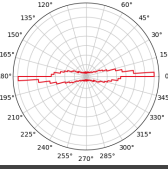
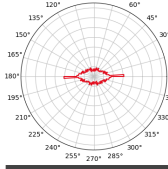
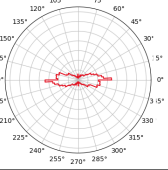
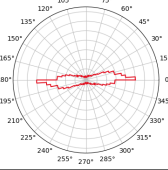
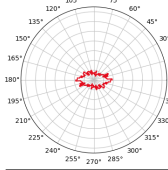
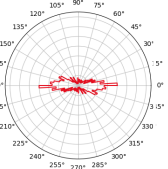
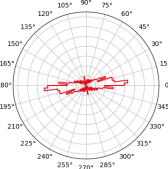
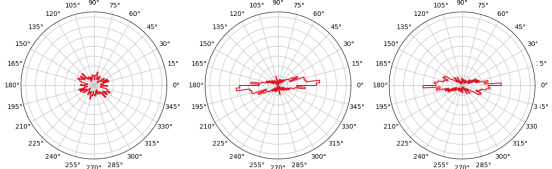
FIBER ORIENTATIONS – SAMPLE B

XY

XZ

YZ

View of a section of the surface in the direction of the X axis.



FIBER IDENTIFICATION ON SAMPLE B

GEODICT

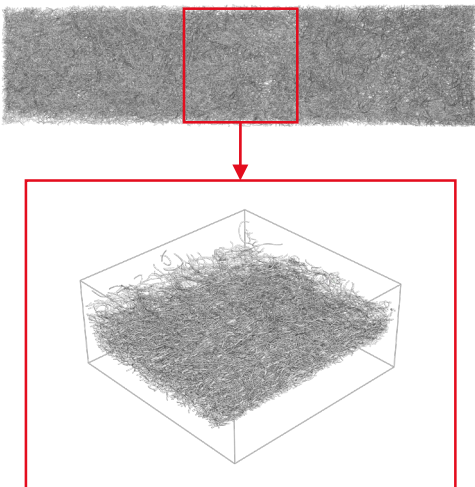
Sample B

Labeling of fibers

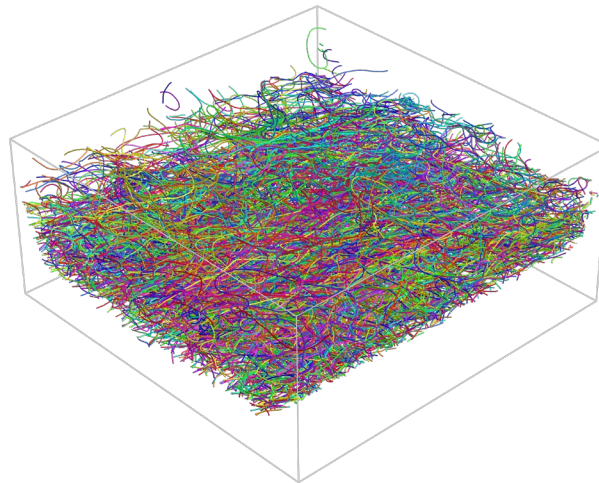
Data becomes information

FiberFind was used on the complete sample.

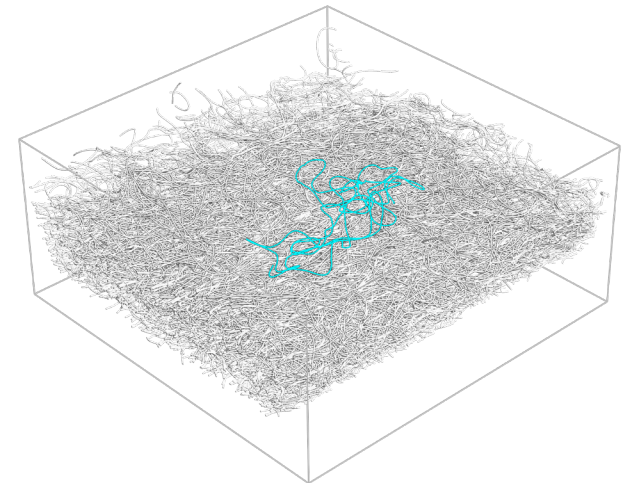
Process is explained on a smaller cutout



The artificial intelligence separates the solid voxels in the image data into individual fibers.
Each fiber becomes an independent, modifiable object which can be treated independently.



Geometric information, such as fiber length, fiber segment orientation and fiber diameter, can be read directly from the object.



- Neural networks separate binder and fibers and identify individual fibers
 - based on learning shapes
- N.N. can label multi-material scans where materials can not be separated by thresholding or classical image processing
- N.N. require training data consisting of segmented and labeled scans
- These can be created easily using material models from GeoDict
 - Models continuously improve, e.g. by improving the binder model
- We continue to improve the capabilities by
 - Increasing the fidelity
 - Speeding up the computations
 - Extending the capabilities to other types of materials
 - Placing the capability to train N.N.s in your hands